## Proof of Taylor's Law for Exponential Growth Models with Migration

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**Presented by Clark Brown** 

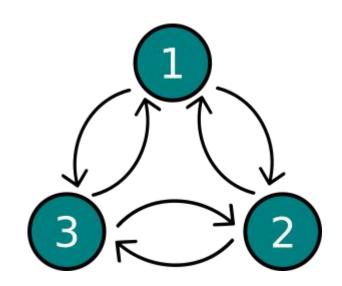
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# An Application of Perron-Frobenius

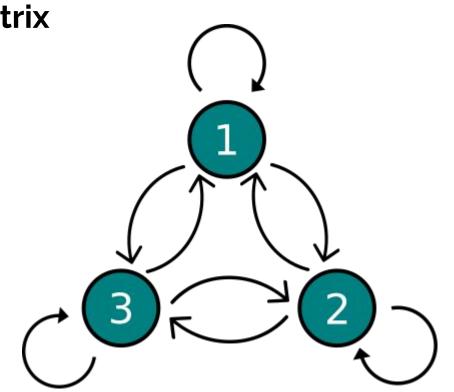
### **Irreducibility of Transition Matrix**

$$M = \begin{bmatrix} -2 & 1 & 1 \\ 1 & -2 & 1 \\ 1 & 1 & -2 \end{bmatrix}$$



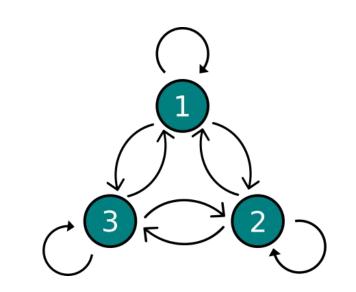
### **Irreducibility of Transition Matrix**

$$3 = \begin{bmatrix} 4 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 4 \end{bmatrix}$$



### **Irreducibility of Transition Matrix**

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



$$\mathbf{N}'(t) = A\mathbf{N}(t)$$

### Perron-Frobenius for Nonnegative Irreducible Matrices

**Theorem.** (Perron-Frobenius Theorem) Let  $A \in \mathbb{R}^{n \times n}$  be nonnegative irreducible matrix. Then (a)  $r = \rho(A)$  is a simple eigenvalue of A; and (b) any eigenvector  $\mathbf{x}$  corresponding to the eigenvalue r has strictly positive entries, i.e.  $\mathbf{x} > 0$ .

Although the Perron-Frobenius theorem is typically stated with more than parts (a) and (b) these are the only results of the theorem we will use to prove the following proposition.

**Proposition.** (Perron-Frobenius for Transition Matrices) Let  $A \in \mathbb{R}^{n \times n}$  be an irreducible transition matrix so that  $a_{ij} \geq 0$  for all  $i \neq j$ . Then there is an eigenvalue  $\lambda_1 \in \sigma(A)$  such that

- (a)  $\lambda_1 \in \mathbb{R}$  is simple;
- (b)  $\lambda_1 > Re(\lambda_i)$  for all  $\lambda_i \in \sigma(A)$  where  $\lambda_i \neq \lambda_1$ ; and
- (c) the eigenvector  $\mathbf{x}_1$  corresponding to  $\lambda_1$  has strictly positive entries, i.e.  $\mathbf{x}_1 > 0$ .

$$A = Q + \operatorname{diag}[m, \dots, m]$$

$$m = \min_{1 < i < n} a_{ii}$$

Q is nonnegative and irreducible, so Perron-Frobenius gives us

$$\lambda \in \sigma(Q) \qquad \exists \ \mathbf{v} \in \mathbb{R}^n, \mathbf{v} \neq \mathbf{0}$$

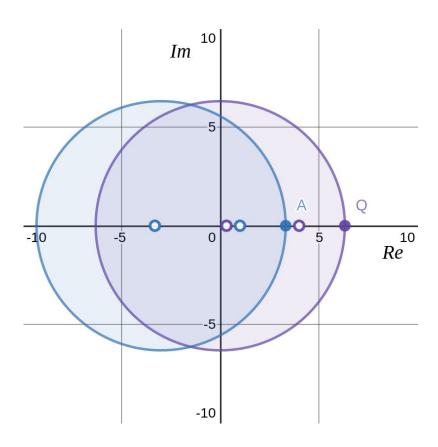
$$Q\mathbf{v} = \lambda \mathbf{v}$$

$$A\mathbf{v} = (Q + mI)\mathbf{v} = \lambda\mathbf{v} + m\mathbf{v} = (\lambda + m)\mathbf{v}$$

$$\lambda + m \in \sigma(A)$$

$$A = \begin{bmatrix} 5 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 5 \end{bmatrix} + \begin{bmatrix} -3 & 0 & 0 \\ 0 & -3 & 0 \\ 0 & 0 & -3 \end{bmatrix}$$

$$\sigma(A) = \left\{ \sqrt{11}, 1, -\sqrt{11} \right\}$$



### **Notation for Proof**

### **General Solution of N(t)**

The spectrum of the transition matrix A is given by:

$$\sigma(A) = \{\lambda_1, \dots, \lambda_m, \alpha_{m+1} + \beta_{m+1}i, \alpha_{m+2} - \beta_{m+2}i, \dots, \alpha_{n-1} + \beta_{n-1}i, \alpha_n - \beta_ni\}$$

$$\overline{\alpha_j + \beta_j i} = \alpha_{j+1} - \beta_{j+1} i$$

With a generalized eigenbasis:

$$\mathbf{x}_1 \dots, \mathbf{x}_n$$

### **General Solution of N(t)**

$$\mathbf{N}(t) = c_1 e^{\lambda_1 t} \mathbf{x}_1 + \sum_{j=2}^{m} c_j t^{p_j} e^{\lambda_j t} \mathbf{x}_j + \sum_{j=m+1}^{n} c_j t^{p_j} e^{\alpha_j t} T_j(\beta_j t) \mathbf{x}_j$$

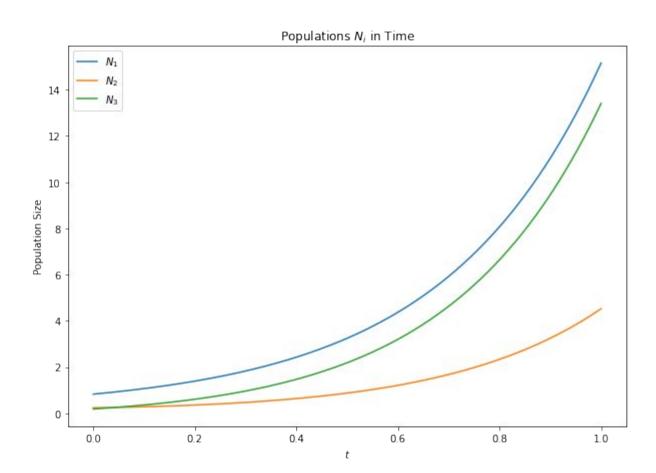
$$T_j(t) = \begin{cases} \cos(t) & \text{for } j = m+1, m+3, \dots, n-1\\ \sin(t) & \text{for } j = m+2, m+4, \dots, n \end{cases}$$

$$c_i \in \mathbb{R}, \ p_i \in \mathbb{Z}^+$$

### **Example Solution of N(t)**

$$\mathbf{N}(t) = c_1 e^{\sqrt{11}t} \mathbf{x}_1 + c_2 e^t \mathbf{x}_2 + c_3 e^{-\sqrt{11}t} \mathbf{x}_3$$

$$\mathbf{x}_1 = \begin{pmatrix} 1 \\ -3 + \sqrt{11} \\ 1 \end{pmatrix}, \mathbf{x}_2 = \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix}, \mathbf{x}_3 = \begin{pmatrix} 1 \\ -3 + \sqrt{11} \\ 1 \end{pmatrix}$$



# Taylor's Law in Exponential Model with Migration

**Definition.** We say that a vector  $\mathbf{v} \in \mathbb{R}^{n \times 1}$  is diagonal if it is a scalar multiple of the vector  $\mathbf{1} = [1 \ 1 \ \dots \ 1]^T$  with components all equal to 1.

**Theorem.** (Sufficient Conditions for Taylor's Law) Let  $A \in \mathbb{R}^{n \times n}$  be a transition matrix and N be the solution to the associated EM model as given previously. If

- (a) the leading coefficient  $c_1 \neq 0$ ;
- (b) the leading eigenvalue  $\lambda_1 \neq 0$ ;
- (c) the leading eigenvector  $\mathbf{x}_1$  is non-diagonal; and
- (d) A is irreducible

then this EM model satisfies Taylor's Law.

# Proof of Taylor's Law in Exponential Model with Migration

### Proof that leading eigenvalue dominates

We write:

$$\mathbf{N}(t) = e^{\lambda_1 t} (c_1 \mathbf{x}_1 + \mathbf{F}(t))$$

$$F_i(t) = \sum_{j=2}^{m} c_j t^{p_j} e^{(\lambda_j - \lambda_1)t} x_{ji} + \sum_{j=m+1}^{m} c_j t^{p_j} e^{(\alpha_j - \lambda_1)t} T_j(\beta_j t) x_{ji}$$

### Proof that leading eigenvalue dominates

$$F_i(t) = \sum_{j=2}^{m} c_j t^{p_j} e^{(\lambda_j - \lambda_1)t} x_{ji} + \sum_{j=m+1}^{m} c_j t^{p_j} e^{(\alpha_j - \lambda_1)t} T_j(\beta_j t) x_{ji}$$

The Transition Matrix A is irreducible, so we have:

$$\lambda_j - \lambda_1 < 0, \ \forall j \neq 1$$

$$\alpha_j - \lambda_1 < 0, \ \forall j \geq m + 1$$

### **Proof that leading eigenvalue dominates**

Hence:

$$\lim_{t \to \infty} F_i(t) = 0$$

Similarly:

$$\lim_{t \to \infty} F_i'(t) = 0$$

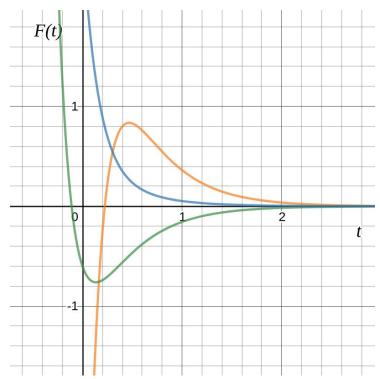
### Proof that leading eigenvalue dominates (EX)

$$\mathbf{N}(t) = e^{\sqrt{11}t} \left( c_1 \mathbf{x}_1 + c_2 e^{(1-\sqrt{11})t} \mathbf{x}_2 + c_3 e^{-2\sqrt{11}t} \mathbf{x}_3 \right)$$

$$F_i(t) = c_2 e^{(1-\sqrt{11})t} x_{2i} + c_3 e^{-2\sqrt{11}t} x_{3i}$$

### **Proof that leading eigenvalue dominates (EX)**

For various values of constants, we see that our F(t) approaches 0 in the limit of large t.



$$b(t) = \frac{d \log[\sum_{i < j} (N_i(t) - N_j(t))^2]}{d \log[\sum_{i=1}^n N_i(t)]}$$

We say Taylor's law holds if we observe the following power-law relationship:

$$\lim_{t \to \infty} b(t) = 2$$

$$b(t) = \frac{d \log[(\sum_{i < j} (N_i(t) - N_j(t))^2]}{d \log[(\sum_{i=1}^n (N_i))]}$$

$$= \frac{d \log[\left(e^{2\lambda_1 t} \sum_{i < j} ((c_1 x_{1i} + F_i(t)) - (c_1 x_{1j} + F_j(t)))^2\right]}{d \log[e^{\lambda_1 t} \left(\sum_{i=1}^n (c_1 x_{1i} + F_i(t))\right]}$$

$$= \frac{d(2\lambda_1 t + \log\left[\sum_{i < j}((c_1 x_{1i} + F_i(t)) - (c_1 x_{1j} + F_j(t)))^2\right])}{d(\lambda_1 t + \log\left[e^{\lambda_1 t}\left(\sum_{i=1}^n(c_1 x_{1i} + F_i(t))\right]\right)}$$

$$= \frac{2\lambda_1 + \frac{2\sum_{i < j}((c_1x_{1i} + F_i(t)) - (c_1x_{1j} + F_j(t)))(F_i'(t) - F_j'(t))}{\sum_{i < j}((c_1x_{1i} + F_i(t)) - (c_1x_{1j} + F_j(t)))^2}}{\lambda_1 + \frac{\sum_{i=1}^n F_i'(t)}{\sum_{i=1}^n (c_1x_{1i} + F_i(t))}}$$

$$\lim_{t \to \infty} b(t) = \frac{2\lambda_1 + \frac{0}{|c_1| \sum_{i < j} |x_{1i} - x_{1j}|^2}}{\lambda_1 + \frac{0}{|c_1| \sum_{i=1}^n |x_{1i}|}}$$

And from the hypotheses, we have:

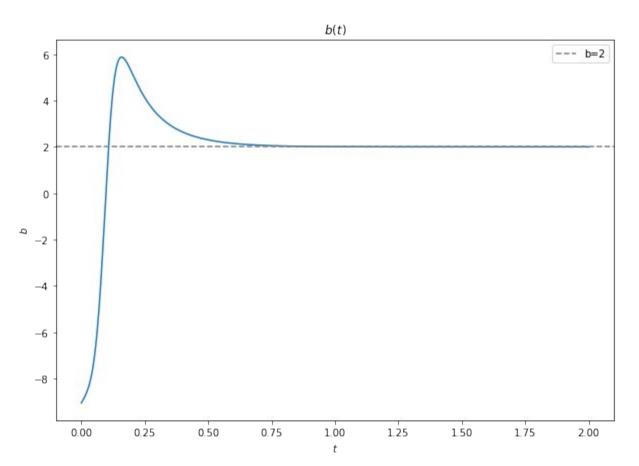
$$\lambda_1 \neq 0, c_1 \neq 0, \exists i, j \text{ such that } x_{1i} \neq x_{1j}$$

$$\lim_{t \to \infty} b(t) = \frac{2\lambda_1}{\lambda_1} = 2$$

So Taylor's Law holds!

$$\lim_{t \to \infty} b(t) = \frac{2\sqrt{11} + \frac{0}{c_1^2 \sum_{i < j} (x_{1i} - x_{1j})^2}}{\sqrt{11} + \frac{0}{c_1 \sum_{i=1}^n x_{1i}}}$$

$$\lim_{t \to \infty} b(t) = \frac{2\sqrt{11}}{\sqrt{11}} = 2$$



## Thank you

### **Q&A**