

Strategy

Metacommunity:

$$\dot{N}_i = rN_i \left[1 - N_i - \sum_{j \neq i} A_{ij} N_j \right] + D[\langle N_i \rangle - N_i] + \sqrt{2T} N_i \eta_i(t),$$

↓ Average

Hierarchy of moments:

$$\langle \dot{N}_i \rangle = \langle N_i \rangle - \langle N_i^2 \rangle - \sum_{j \neq i} A_{ij} \langle N_i N_j \rangle$$

Strategy

Metacommunity:

$$\dot{N}_i = rN_i \left[1 - N_i - \sum_{j \neq i} A_{ij} N_j \right] + D[\langle N_i \rangle - N_i] + \sqrt{2T} N_i \eta_i(t),$$

↓ Average

Hierarchy of moments:

$$\langle \dot{N}_i \rangle = \langle N_i \rangle - \langle N_i^2 \rangle - \sum_{j \neq i} A_{ij} \langle N_i N_j \rangle$$

↓ $D \gg r$

Integrate out fluctuations:

$$\langle \dot{N}_i \rangle = f_i(\langle N_1 \rangle, \dots, \langle N_S \rangle)$$

Strategy

Metacommunity:

$$\dot{N}_i = rN_i \left[1 - N_i - \sum_{j \neq i} A_{ij} N_j \right] + D[\langle N_i \rangle - N_i] + \sqrt{2T} N_i \eta_i(t),$$

↓ Average

Hierarchy of moments:

$$\langle \dot{N}_i \rangle = \langle N_i \rangle - \langle N_i^2 \rangle - \sum_{j \neq i} A_{ij} \langle N_i N_j \rangle$$

↓ $D \gg r$

Integrate out fluctuations:

$$\langle \dot{N}_i \rangle = f_i(\langle N_1 \rangle, \dots, \langle N_S \rangle)$$

$T < D$

GLV

(low diversity)

$T > D$

θ -GLV

(high diversity iff $T > 2D$)

The $D \gg r$ adiabatic limit

$$\dot{N}_i = rN_i[1 - N_i - \sum_{j \neq i} A_{ij}N_j] + D[\langle N_i \rangle - N_i] + \sqrt{2T}N_i\eta_i(t),$$

The $D \gg r$ adiabatic limit

$$\dot{N}_i = rN_i[1 - N_i - \sum_{j \neq i} A_{ij}N_j] + D[\langle N_i \rangle - N_i] + \sqrt{2T}N_i\eta_i(t),$$

- Average over patches:

$$\langle \dot{N}_i \rangle = r \left[\langle N_i \rangle - \langle N_i^2 \rangle - \sum_{j \neq i} A_{ij} \langle N_i N_j \rangle \right],$$

The $D \gg r$ adiabatic limit

$$\dot{N}_i = rN_i[1 - N_i - \sum_{j \neq i} A_{ij}N_j] + D[\langle N_i \rangle - N_i] + \sqrt{2T}N_i\eta_i(t),$$

- Average over patches:

$$\langle \dot{N}_i \rangle = r \left[\langle N_i \rangle - \langle N_i^2 \rangle - \sum_{j \neq i} A_{ij} \langle N_i N_j \rangle \right],$$

- Hierarchy of moments:

$$\begin{aligned}\partial_t \langle N_i^2 \rangle &= 2D \langle N_i \rangle^2 - 2(D - T) \langle N_i^2 \rangle + 2r [\langle N_i^2 \rangle - \langle N_i^3 \rangle - \sum_j A_{ij} \langle N_i^2 N_j \rangle], \\ \partial_t \langle N_i N_j \rangle &= 2D (\langle N_i \rangle \langle N_j \rangle - \langle N_i N_j \rangle) + \mathcal{O}(r),\end{aligned}$$

The $D \gg r$ adiabatic limit

$$\dot{N}_i = rN_i[1 - N_i - \sum_{j \neq i} A_{ij}N_j] + D[\langle N_i \rangle - N_i] + \sqrt{2T}N_i\eta_i(t),$$

- Average over patches:

$$\langle \dot{N}_i \rangle = r \left[\langle N_i \rangle - \langle N_i^2 \rangle - \sum_{j \neq i} A_{ij} \langle N_i N_j \rangle \right],$$

- Hierarchy of moments:

$$\begin{aligned}\partial_t \langle N_i^2 \rangle &= 2D \langle N_i \rangle^2 - 2(D - T) \langle N_i^2 \rangle + 2r [\langle N_i^2 \rangle - \langle N_i^3 \rangle - \sum_j A_{ij} \langle N_i^2 N_j \rangle], \\ \partial_t \langle N_i N_j \rangle &= 2D (\langle N_i \rangle \langle N_j \rangle - \langle N_i N_j \rangle) + \mathcal{O}(r),\end{aligned}$$

- For $D, T \gg r$, second moments enslaved to slow-evolving means

The $D \gg r$ adiabatic limit

$$\dot{N}_i = rN_i[1 - N_i - \sum_{j \neq i} A_{ij}N_j] + D[\langle N_i \rangle - N_i] + \sqrt{2T}N_i\eta_i(t),$$

- Average over patches:

$$\langle \dot{N}_i \rangle = r \left[\langle N_i \rangle - \langle N_i^2 \rangle - \sum_{j \neq i} A_{ij} \langle N_i N_j \rangle \right],$$

- Hierarchy of moments:

$$\begin{aligned}\partial_t \langle N_i^2 \rangle &= 2D \langle N_i \rangle^2 - 2(D - T) \langle N_i^2 \rangle + 2r [\langle N_i^2 \rangle - \langle N_i^3 \rangle - \sum_j A_{ij} \langle N_i^2 N_j \rangle], \\ \partial_t \langle N_i N_j \rangle &= 2D (\langle N_i \rangle \langle N_j \rangle - \langle N_i N_j \rangle) + \mathcal{O}(r),\end{aligned}$$

- For $D, T \gg r$, second moments enslaved to slow-evolving means:

$$\langle N_i N_{j \neq i} \rangle = \langle N_i \rangle \langle N_j \rangle + \mathcal{O}\left(\frac{r}{D}\right), \quad \langle N_i^2 \rangle = \frac{D}{D-T} \langle N_i \rangle^2 + \mathcal{O}\left(\frac{r}{D-T}\right).$$

The $D \gg r$ adiabatic limit

$$\dot{N}_i = rN_i[1 - N_i - \sum_{j \neq i} A_{ij}N_j] + D[\langle N_i \rangle - N_i] + \sqrt{2T}N_i\eta_i(t),$$

- Average over patches:

$$\langle \dot{N}_i \rangle = r \left[\langle N_i \rangle - \langle N_i^2 \rangle - \sum_{j \neq i} A_{ij} \langle N_i N_j \rangle \right],$$

- Hierarchy of moments:

$$\begin{aligned}\partial_t \langle N_i^2 \rangle &= 2D \langle N_i \rangle^2 - 2(D - T) \langle N_i^2 \rangle + 2r [\langle N_i^2 \rangle - \langle N_i^3 \rangle - \sum_j A_{ij} \langle N_i^2 N_j \rangle], \\ \partial_t \langle N_i N_j \rangle &= 2D (\langle N_i \rangle \langle N_j \rangle - \langle N_i N_j \rangle) + \mathcal{O}(r),\end{aligned}$$

- For $D, T \gg r$, second moments enslaved to slow-evolving means:

$$\langle N_i N_{j \neq i} \rangle = \langle N_i \rangle \langle N_j \rangle + \mathcal{O}\left(\frac{r}{D}\right), \quad \langle N_i^2 \rangle = \frac{D}{D-T} \langle N_i \rangle^2 + \mathcal{O}\left(\frac{r}{D-T}\right).$$

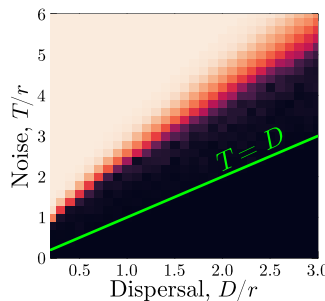
Valid whenever $D \gg r$

Valid only for $T < D$

Effective GLV dynamics for $T < D$

- Effective equation for means ($D - T \gg r$):

$$\partial_t \langle N_i \rangle = r \langle N_i \rangle \left[1 - \frac{D}{D - T} \langle N_i \rangle - \sum_{j \neq i} A_{ij} \langle N_j \rangle \right]$$



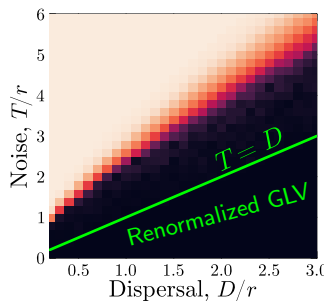
Effective GLV dynamics for $T < D$

- **Effective equation for means** ($D - T \gg r$):

$$\partial_t \langle N_i \rangle = r \langle N_i \rangle \left[1 - \frac{D}{D - T} \langle N_i \rangle - \sum_{j \neq i} A_{ij} \langle N_j \rangle \right]$$

$\Rightarrow$ *Standard GLV with renormalized carrying capacity*

$\Rightarrow S^* \in \mathcal{O}(1), \phi = 0 \checkmark$



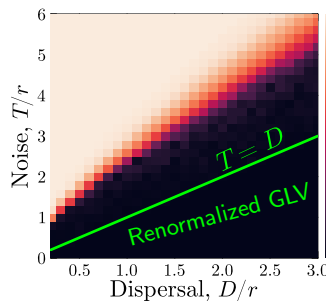
Effective GLV dynamics for $T < D$

- Effective equation for means ($D - T \gg r$):

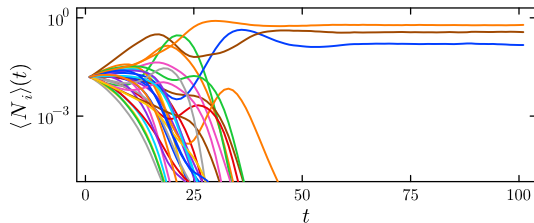
$$\partial_t \langle N_i \rangle = r \langle N_i \rangle \left[1 - \frac{D}{D - T} \langle N_i \rangle - \sum_{j \neq i} A_{ij} \langle N_j \rangle \right]$$

$\Rightarrow$ Standard GLV with renormalized carrying capacity

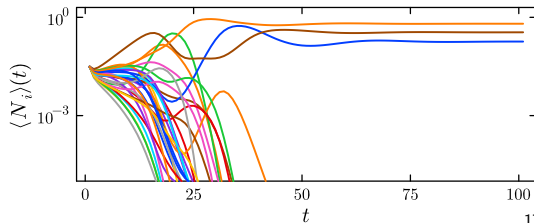
$\Rightarrow S^* \in \mathcal{O}(1), \phi = 0 \checkmark$



Metacommunity ($D = 1, T = 0.1$):



Effective GLV dynamics:



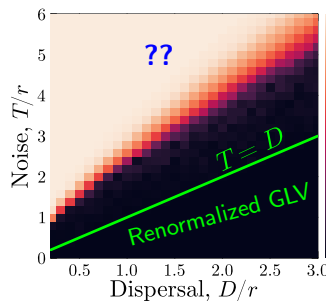
Effective GLV dynamics for $T < D$

- Effective equation for means ($D - T \gg r$):

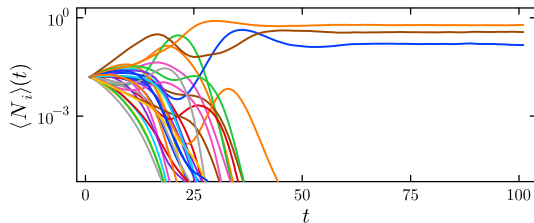
$$\partial_t \langle N_i \rangle = r \langle N_i \rangle \left[1 - \frac{D}{D - T} \langle N_i \rangle - \sum_{j \neq i} A_{ij} \langle N_j \rangle \right]$$

$\Rightarrow$ Standard GLV with renormalized carrying capacity

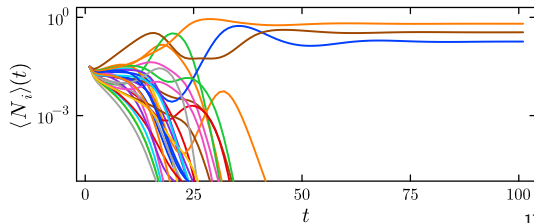
$\Rightarrow S^* \in \mathcal{O}(1), \phi = 0 \checkmark$



Metacommunity ($D = 1, T = 0.1$):



Effective GLV dynamics:



Stationary abundance distribution in coexistence phase

$$\dot{N}_i = N_i [1 - N_i - \xi_i] + D[\langle N_i \rangle - N_i] + \sqrt{2T} N_i \eta_i(t), \quad \xi_i(t) \equiv \sum_{j \neq i} A_{ij} N_j(t)$$

Stationary abundance distribution in coexistence phase

$$\dot{N}_i = N_i [1 - N_i - \xi_i] + D[\langle N_i \rangle - N_i] + \sqrt{2T} N_i \eta_i(t), \quad \xi_i(t) \equiv \sum_{j \neq i} A_{ij} N_j(t)$$

► Coexistence phase: Assume $\langle N_i \rangle > 0$, $\langle N_i \rangle \in \mathcal{O}(1/S)$

Stationary abundance distribution in coexistence phase

$$\dot{N}_i = N_i [1 - N_i - \xi_i] + D[\langle N_i \rangle - N_i] + \sqrt{2T} N_i \eta_i(t), \quad \xi_i(t) \equiv \sum_{j \neq i} A_{ij} N_j(t)$$

► **Coexistence phase:** Assume $\langle N_i \rangle > 0$, $\langle N_i \rangle \in \mathcal{O}(1/S)$

► **Claim:** for $D \gg r$, fluctuations of ξ_i vanish at large S :

$$\xi_i(t) \xrightarrow{S \rightarrow \infty} \langle \xi_i \rangle \equiv \sum_{j \neq i} A_{ij} \langle N_j \rangle$$

To show this:

1. Assume $\xi = \langle \xi \rangle$, find $\langle \xi \rangle$ self-consistently
2. Compute $\langle N_i^2 \rangle$
3. Show $\lim_{S \rightarrow \infty} \langle \xi^2 \rangle_c = 0$

Stationary abundance distribution in coexistence phase

$$\dot{N}_i = N_i [1 - N_i - \langle \xi_i \rangle] + D[\langle N_i \rangle - N_i] + \sqrt{2T} N_i \eta_i(t)$$

Stationary abundance distribution in coexistence phase

$$\dot{N}_i = N_i [1 - N_i - \langle \xi_i \rangle] + D[\langle N_i \rangle - N_i] + \sqrt{2T} N_i \eta_i(t)$$

- Stationary probability density ($\beta \equiv 1/T$):

$$p_i(N) \propto e^{-\beta N} e^{-\beta D \langle N_i \rangle / N} N^{-\gamma_i}, \quad \gamma_i \equiv 2 + \beta(D + \langle \xi_i \rangle - 1).$$

Stationary abundance distribution in coexistence phase

$$\dot{N}_i = N_i [1 - N_i - \langle \xi_i \rangle] + D[\langle N_i \rangle - N_i] + \sqrt{2T} N_i \eta_i(t)$$

- Stationary probability density ($\beta \equiv 1/T$):

$$p_i(N) \propto e^{-\beta N} e^{-\beta D \langle N_i \rangle / N} N^{-\gamma_i}, \quad \gamma_i \equiv 2 + \beta(D + \langle \xi_i \rangle - 1).$$

- Impose self-consistency:

$$\langle N_i \rangle = \int_0^\infty dN N p_i(N) = \frac{\sqrt{D \langle N_i \rangle} K_{\gamma-2}(2\beta \sqrt{D \langle N_i \rangle})}{K_{\gamma-1}(2\beta \sqrt{D \langle N_i \rangle})}.$$

Stationary abundance distribution in coexistence phase

$$\dot{N}_i = N_i [1 - N_i - \langle \xi_i \rangle] + D[\langle N_i \rangle - N_i] + \sqrt{2T} N_i \eta_i(t)$$

- Stationary probability density ($\beta \equiv 1/T$):

$$p_i(N) \propto e^{-\beta N} e^{-\beta D \langle N_i \rangle / N} N^{-\gamma_i}, \quad \gamma_i \equiv 2 + \beta(D + \langle \xi_i \rangle - 1).$$

- Impose self-consistency:

$$\langle N_i \rangle = \int_0^\infty dN N p_i(N) = \frac{\sqrt{D \langle N_i \rangle} K_{\gamma-2}(2\beta \sqrt{D \langle N_i \rangle})}{K_{\gamma-1}(2\beta \sqrt{D \langle N_i \rangle})}.$$

- Expand for small $\langle N_i \rangle \propto 1/S$ and solve for $\langle \xi \rangle$:

$$\boxed{\langle \xi_i \rangle = 1 + \mathcal{O}(S^{-\beta D})} \text{ for all } i.$$

Power law distribution of abundances

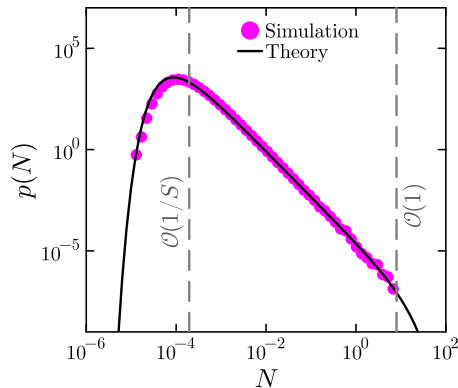
► Result:

All species follow the same power law

$$p_i(N) \propto e^{-\beta N} e^{-\beta D \langle N_i \rangle / N} N^{-2-\beta D},$$

► Power laws observed in natural ecosystems:

(Locey and Lennon, 2016; Ser-Giacomi, 2018; etc.)



Power law distribution of abundances

► Result:

All species follow the same power law

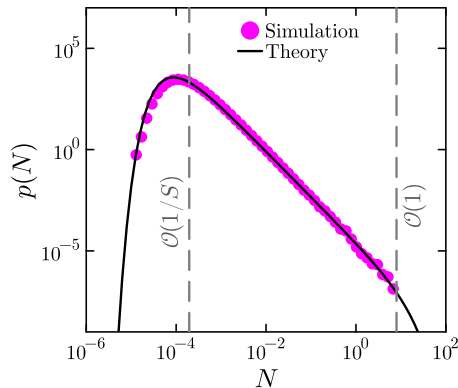
$$p_i(N) \propto e^{-\beta N} e^{-\beta D \langle N_i \rangle / N} N^{-2-\beta D},$$

► Power laws observed in natural ecosystems:

(Locey and Lennon, 2016; Ser-Giacomi, 2018; etc.)

► Consequence:

Anomalous scaling of moments



Anomalous scaling of moments

- For small $\langle N \rangle \propto 1/S$:

$$\langle N^2 \rangle \simeq \frac{D}{D-T} \langle N \rangle^2 \left[1 + (\dots) \langle N \rangle^{\beta D - 1} \right] + \mathcal{O}(\langle N \rangle^2)$$

Anomalous scaling of moments

- For small $\langle N \rangle \propto 1/S$:

$$\langle N^2 \rangle \simeq \frac{D}{D-T} \langle N \rangle^2 \left[1 + (\dots) \langle N \rangle^{\beta D - 1} \right] + \mathcal{O}(\langle N \rangle^2)$$

- Transition at $T = D$ (Swartz et al. 2022, Ottino-Löffler & Kardar 2020):

$$\langle N_i^2 \rangle \propto \begin{cases} \langle N_i \rangle^2, & T < D, \\ \langle N_i \rangle^{1+\beta D}, & T > D. \end{cases}$$

Anomalous scaling of moments

- For small $\langle N \rangle \propto 1/S$:

$$\langle N^2 \rangle \simeq \frac{D}{D-T} \langle N \rangle^2 \left[1 + (\dots) \langle N \rangle^{\beta D - 1} \right] + \mathcal{O}(\langle N \rangle^2)$$

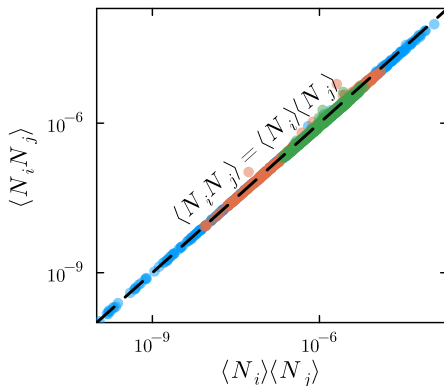
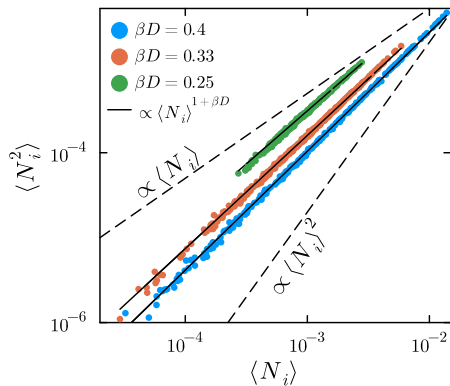
- Transition at $T = D$ (Swartz et al. 2022, Ottino-Löffler & Kardar 2020):

$$\langle N_i^2 \rangle \propto \begin{cases} \langle N_i \rangle^2, & T < D, \\ \langle N_i \rangle^{1+\beta D}, & T > D. \end{cases}$$

- **Taylor's power law (1961):** $\langle N_i^2 \rangle_c \propto \langle N_i \rangle^{1+\theta}$

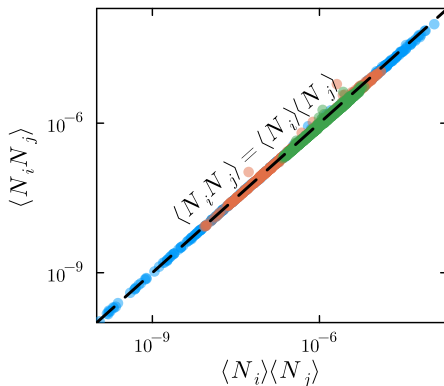
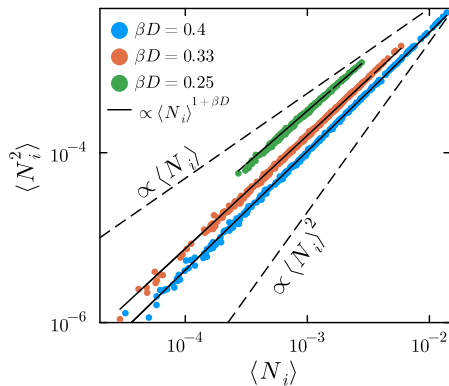
Widely verified in ecological data, nonuniversal exponent (Taylor 1961; Eisler et al. 2008)

Anomalous scaling of moments in the coexistence phase



Simulations support $\langle N_i^2 \rangle \propto \langle N_i \rangle^{1+\beta D}$ in coexistence phase.

Anomalous scaling of moments in the coexistence phase



Simulations support $\langle N_i^2 \rangle \propto \langle N_i \rangle^{1+\beta D}$ in coexistence phase.

$$\Rightarrow \boxed{\langle N_i \rangle - \langle N_i^2 \rangle - \sum_j A_{ij} \langle N_i N_j \rangle} \xrightarrow{D \gg r} \boxed{\langle N_i \rangle - \langle N_i \rangle^{1+\theta} - \sum_j A_{ij} \langle N_i \rangle \langle N_j \rangle}$$

θ -logistic Generalized Lotka-Volterra Equation

- Effective equation for means ($\theta \equiv \beta D$)

$$\langle \dot{N}_i \rangle = \langle N_i \rangle \left[1 - \langle N_i \rangle^\theta - \sum_{j \neq i} A_{ij} \langle N_j \rangle \right], \quad (\theta\text{-GLV}).$$

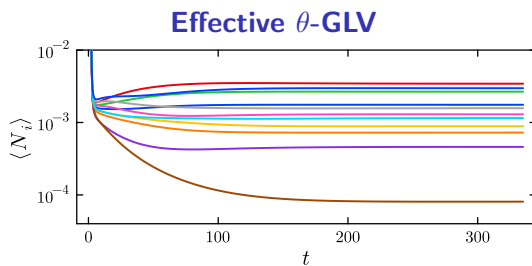
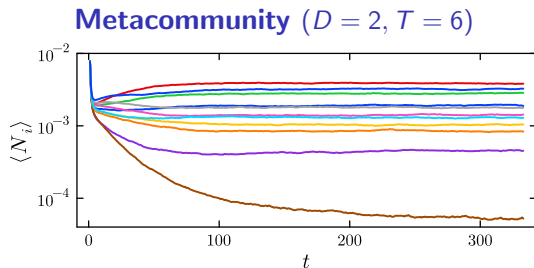
- Sometimes called **θ -logistic** self-regulation (Hatton et al.).

θ -logistic Generalized Lotka-Volterra Equation

- Effective equation for means ($\theta \equiv \beta D$)

$$\langle \dot{N}_i \rangle = \langle N_i \rangle \left[1 - \langle N_i \rangle^\theta - \sum_{j \neq i} A_{ij} \langle N_j \rangle \right], \quad (\theta\text{-GLV}).$$

- Sometimes called **θ -logistic** self-regulation (Hatton et al.).

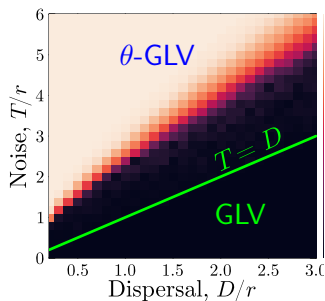


θ -logistic Generalized Lotka-Volterra Equation

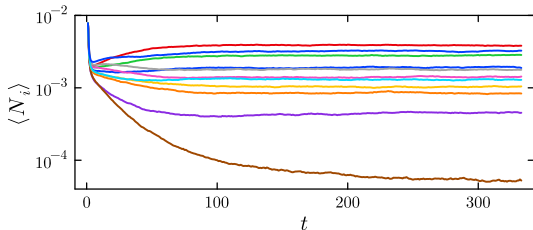
- Effective equation for means ($\theta \equiv \beta D$)

$$\langle \dot{N}_i \rangle = \langle N_i \rangle \left[1 - \langle N_i \rangle^\theta - \sum_{j \neq i} A_{ij} \langle N_j \rangle \right], \quad (\theta\text{-GLV}).$$

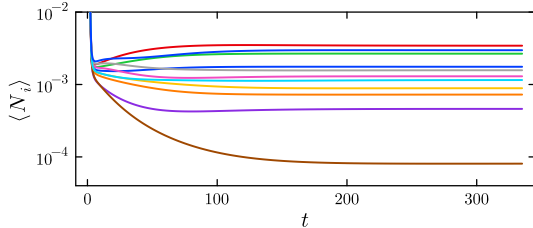
- Sometimes called **θ -logistic** self-regulation (Hatton et al.).



Metacommunity ($D = 2, T = 6$)



Effective θ -GLV



Fixed points of the θ -GLV model

$$0 = \langle N_i \rangle \left[1 - \langle N_i \rangle^\theta - \sum_{j \neq i} A_{ij} \langle N_j \rangle \right], \quad \overline{A_{ij}} = \mu, \quad \overline{A_{ij} A_{kl}} = \sigma^2 \delta_{ik} \delta_{jl}.$$

- Two solutions: $\langle N_i \rangle = 0$ or $\langle N_i \rangle = (1 - \sum_j A_{ij} \langle N_j \rangle)^{1/\theta}$

Fixed points of the θ -GLV model

$$0 = \langle N_i \rangle \left[1 - \langle N_i \rangle^\theta - \sum_{j \neq i} A_{ij} \langle N_j \rangle \right], \quad \overline{A_{ij}} = \mu, \quad \overline{A_{ij} A_{kl}} = \sigma^2 \delta_{ik} \delta_{jl}.$$

- ▶ Two solutions: $\langle N_i \rangle = 0$ or $\langle N_i \rangle = (1 - \sum_j A_{ij} \langle N_j \rangle)^{1/\theta}$
- ▶ Cavity method: replace $\sum_{j \neq i} A_{ij} \langle N_j \rangle$ with mean + Gaussian fluctuations:

$$\langle N \rangle = \max \left(0, 1 - \mu S \overline{\langle N \rangle} - \sigma \sqrt{S \overline{\langle N \rangle}^2} Z \right)^{1/\theta}, \quad Z \sim \mathcal{N}(0, 1).$$

with $\overline{\langle N \rangle}, \overline{\langle N \rangle}^2$ determined self-consistently.

Fixed points of the θ -GLV model

$$0 = \langle N_i \rangle \left[1 - \langle N_i \rangle^\theta - \sum_{j \neq i} A_{ij} \langle N_j \rangle \right], \quad \overline{A_{ij}} = \mu, \quad \overline{A_{ij} A_{kl}} = \sigma^2 \delta_{ik} \delta_{jl}.$$

- ▶ Two solutions: $\langle N_i \rangle = 0$ or $\langle N_i \rangle = (1 - \sum_j A_{ij} \langle N_j \rangle)^{1/\theta}$
- ▶ Cavity method: replace $\sum_{j \neq i} A_{ij} \langle N_j \rangle$ with mean + Gaussian fluctuations:

$$\langle N \rangle = \max \left(0, 1 - \mu S \overline{\langle N \rangle} - \sigma \sqrt{S \overline{\langle N \rangle}^2} Z \right)^{1/\theta}, \quad Z \sim \mathcal{N}(0, 1).$$

with $\overline{\langle N \rangle}, \overline{\langle N \rangle}^2$ determined self-consistently.

- ▶ Survival fraction $\phi = \mathbb{P}[\langle N \rangle > 0]$

Fixed points of the θ -GLV model

$$0 = \langle N_i \rangle \left[1 - \langle N_i \rangle^\theta - \sum_{j \neq i} A_{ij} \langle N_j \rangle \right], \quad \overline{A_{ij}} = \mu, \quad \overline{A_{ij} A_{kl}} = \sigma^2 \delta_{ik} \delta_{jl}.$$

- ▶ Two solutions: $\langle N_i \rangle = 0$ or $\langle N_i \rangle = (1 - \sum_j A_{ij} \langle N_j \rangle)^{1/\theta}$
- ▶ Cavity method: replace $\sum_{j \neq i} A_{ij} \langle N_j \rangle$ with mean + Gaussian fluctuations:

$$\langle N \rangle = \max \left(0, 1 - \mu S \overline{\langle N \rangle} - \sigma \sqrt{S \overline{\langle N \rangle}^2} Z \right)^{1/\theta}, \quad Z \sim \mathcal{N}(0, 1).$$

with $\overline{\langle N \rangle}, \overline{\langle N \rangle}^2$ determined self-consistently.

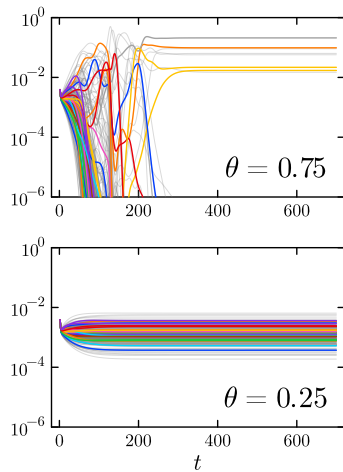
- ▶ Survival fraction $\phi = \mathbb{P}[\langle N \rangle > 0]$
- ▶ Solve using large- S ansatz: $\overline{\langle N \rangle} \sim S^{-\alpha}, \overline{\langle N \rangle}^2 \sim S^{-\gamma}, \phi \rightarrow 1$

Cavity solution of the θ -GLV model

$$0 = \langle N_i \rangle \left[1 - \langle N_i \rangle^\theta - \sum_{j \neq i} A_{ij} \langle N_j \rangle \right]$$

Results:

- There is a coexistence solution ($\phi = 1$) iff $\theta < \frac{1}{2}$

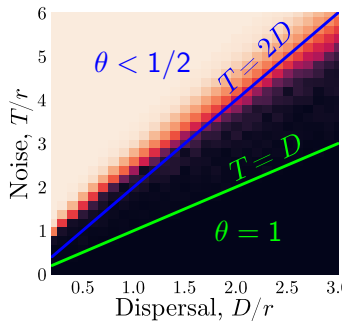


Cavity solution of the θ -GLV model

$$0 = \langle N_i \rangle \left[1 - \langle N_i \rangle^\theta - \sum_{j \neq i} A_{ij} \langle N_j \rangle \right]$$

Results:

- **There is a coexistence solution ($\phi = 1$) iff $\theta < \frac{1}{2}$**
Using $\theta = \beta D$, this implies $T_c = 2D$



Cavity solution of the θ -GLV model

$$0 = \langle N_i \rangle \left[1 - \langle N_i \rangle^\theta - \sum_{j \neq i} A_{ij} \langle N_j \rangle \right]$$

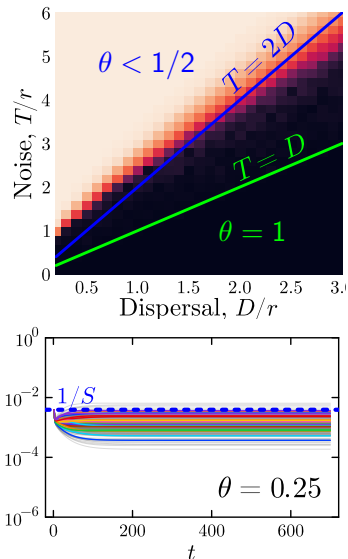
Results:

- **There is a coexistence solution ($\phi = 1$) iff $\theta < \frac{1}{2}$**
Using $\theta = \beta D$, this implies $T_c = 2D$

- In coexistence phase:

$$\overline{\langle N \rangle} \simeq \frac{1}{\mu S}, \quad \overline{\langle N \rangle^2} \simeq \frac{1}{\mu^2 S^2}.$$

$\Rightarrow$ **mean abundances** $\in \mathcal{O}(1/S)$.



Cavity solution of the θ -GLV model

$$0 = \langle N_i \rangle \left[1 - \langle N_i \rangle^\theta - \sum_{j \neq i} A_{ij} \langle N_j \rangle \right]$$

Results:

- **There is a coexistence solution ($\phi = 1$) iff $\theta < \frac{1}{2}$**
Using $\theta = \beta D$, this implies **$T_c = 2D$**

- In coexistence phase:

$$\overline{\langle N \rangle} \simeq \frac{1}{\mu S}, \quad \overline{\langle N \rangle^2} \simeq \frac{1}{\mu^2 S^2}.$$

$\Rightarrow$ **mean abundances** $\in \mathcal{O}(1/S)$.

- As $S \rightarrow \infty$, $\overline{\langle N \rangle^2} \rightarrow \overline{\langle N \rangle}^2$
 $\Rightarrow$ **disorder becomes irrelevant.**

