

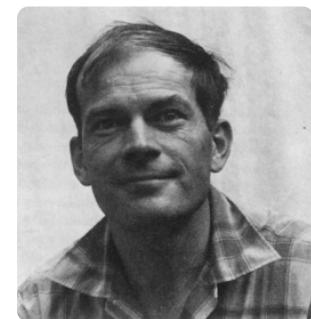
B2. Stability in generalized CR model

Recall generalized Lotka-Volterra model:

$$\dot{p}_i = (r_i - \sum_j A_{ij} p_j) p_i$$

Many-Species CR Model

(Robert MacArthur, 1970)



$$\dot{p}_i = \sum_{\alpha=1}^{N_R} V_{i\alpha} n_\alpha p_i - \mu_i p_i \quad N_R = \# \text{"resources"} \\ N_C = \# \text{"consumers"}$$

$$\dot{n}_\alpha = Y_\alpha n_\alpha (1 - n_\alpha / K_\alpha) - \sum_{i=1}^{N_C} V_{i\alpha} n_\alpha p_i / Y_\alpha$$

Compared to the CR model in Sec. IB1,

$$\dot{p}_i = \left[\sum_\alpha V_{i\alpha} n_\alpha - \mu \right] p_i$$

$$\dot{n}_\alpha = \mu (n_\alpha^0 - n_\alpha) - \sum_i V_{i\alpha} n_\alpha p_i / Y_\alpha$$

Nutrient in MacArthur's model "self-generated,"

$$Y_\alpha n_\alpha \leftrightarrow \mu n_\alpha^0, \quad K_\alpha \leftrightarrow n_\alpha^0$$

- makes mathematics simpler
- no major diff. except for dependence on specific parameters
(See Butler & O'Dwyer, 2018)

MacArthur showed feasible soln ($p_i^* > 0, n_\alpha^* > 0$) are global attractor of CR dynamics as long as $N_c \leq N_R$, i.e., \leq one species/niche = "ecological exclusion principle"

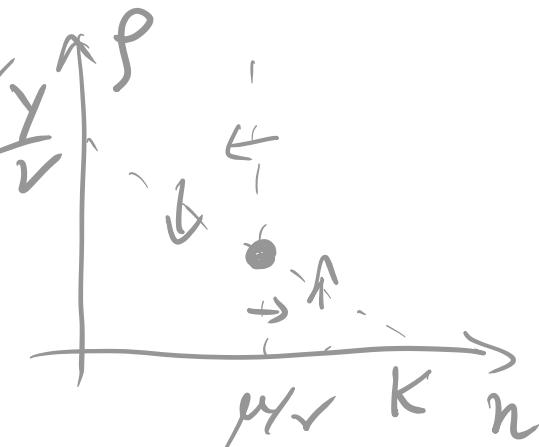
→ We reconstruct below MacArthur's work.

- Recall predator-prey dynamics with carrying cap. (See IA3b)

$$\begin{aligned} \dot{n} &= \gamma n (1 - n/K) - \nu n p / \gamma & n = \text{prey} \\ \dot{p} &= \gamma n p - \mu p & p = \text{predator} \end{aligned}$$

Steady-state: $\gamma n^* = \mu$

$$\gamma (1 - n^*/K) = \nu p^* / \gamma$$



→ fixed point $p^* = \frac{\gamma Y}{\nu} (1 - \mu/\nu K)$
is stable if $\mu/\nu < K$ ($p^* > 0$)

General N_R, N_c :

fixed pt $p_i^* = \sum_j A_{ij}^{-1} r_j$ is stable

with $r_i = \sum_\alpha v_{i\alpha} k_\alpha - \mu_i$, $A_{ij} = \sum_\alpha v_{i\alpha} v_{j\alpha} k_\alpha / \gamma_\alpha Y_\alpha$

if $N_R \geq N_c$: competitive exclusion

$p_i^* > 0$: stability = feasibility

a) Solve for fixed pt of many-species system

$$\dot{n}_\alpha = 0 \rightarrow Y_\alpha \left(1 - \frac{n_\alpha}{K_\alpha} \right) = \sum_i \nu_{i\alpha} f_i / Y_\alpha$$

$$\begin{matrix} \checkmark \\ n_\alpha \neq 0 \end{matrix} \quad n_\alpha = K_\alpha - \sum_i \nu_{i\alpha} f_i \underbrace{\left(K_\alpha / Y_\alpha \right)}_{\sigma_\alpha}$$

Note: MacArt assumed nutrient to relax with faster time scale; full system studied by Case & Casten (1979).

Substitute into eqn for f_i :

$$\dot{f}_i = \sum_\alpha \nu_{i\alpha} f_i n_\alpha - \mu_i f_i$$

$$= \sum_\alpha \nu_{i\alpha} f_i \left(K_\alpha - \sum_j \nu_{j\alpha} f_j \sigma_\alpha \right) - \mu_i f_i$$

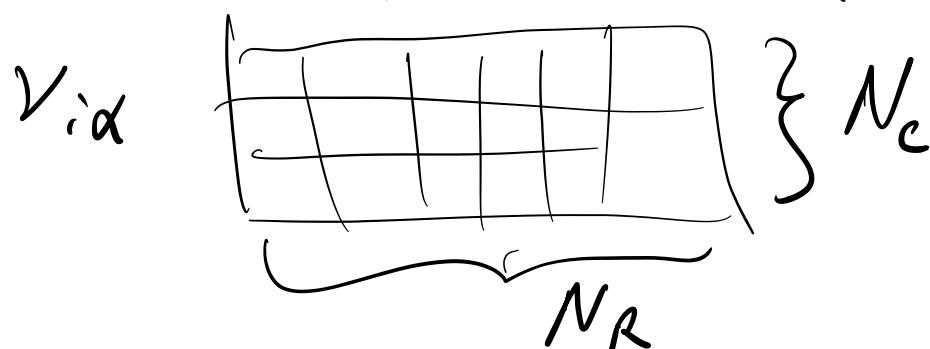
$$\dot{f}_i = \underbrace{\left(\sum_\alpha \nu_{i\alpha} K_\alpha - \mu_i \right)}_{r_i} f_i - \sum_{j\alpha} \nu_{i\alpha} \nu_{j\alpha} \sigma_\alpha f_i f_j$$

\rightarrow effective gLV system

Note: f_i , $i = 1, \dots, N_c$

n_α , $\alpha = 1, \dots, N_R$

(will see generic
Solv requires
 $N_c \leq N_R$)



$A_{ij} = \sum_\alpha \nu_{i\alpha} \nu_{j\alpha} \sigma_\alpha$ is $N_c \times N_c$ (outer product)

$$\text{Steady State: } r_i = \sum_j A_{ij} s_j$$

$$\rightarrow s_i^* = \sum_j A_{ij}^{-1} r_j$$

(provided that A_{ij} invertible
- see below)

b) Stability of fixed pt:

Next show that $s_i^* = \sum_j A_{ij}^{-1} r_j$

is the global fixed point of the glV eqn:

$$\frac{ds_i}{dt} = r_i s_i - \sum_j A_{ij} s_i s_j$$

$$\text{where } A_{ij} = \sum_\alpha V_{ia} V_{ja} \sigma_\alpha$$

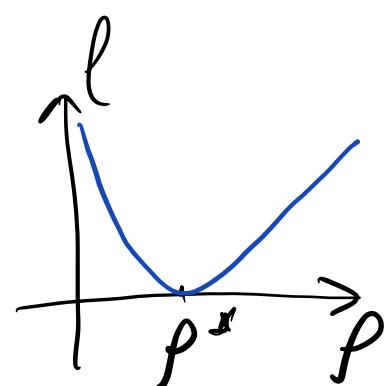
→ Construct "Lyapunov function":

$$L(t) = \sum_i \underbrace{\left[(s_i(t) - s_i^*) - s_i^* \ln \frac{s_i(t)}{s_i^*} \right]}_{l_i(t)}$$

Property of each term l_i :

$$l_i(t)$$

$$\begin{aligned} \frac{dl}{dp} &= 1 - \frac{p^*}{p} \\ \frac{dl}{dp^2} &= \frac{p^*}{p^2} > 0 \end{aligned} \quad \left\{ \begin{array}{l} \frac{dl}{dp} = 0 \text{ at } p = p^* \\ l(p^*) = 0 \end{array} \right.$$



$\rightarrow \ell(p)$ has single minimum at p^*

$\therefore L = \sum_i l_i$ has one global min at $p_i = p_i^*$
for $p_i > 0, p_i^* > 0$

Next, look at $\frac{dL}{dt}$

$$\frac{dL(t)}{dt} = \sum_i \left[\frac{dp_i}{dt} - \frac{p_i^*}{p_i} \frac{df_i}{dt} \right]$$

use ∇f_i for $\frac{df_i}{dt}$ \Downarrow

$$= \sum_i \left[(r_i - \sum_j A_{ij} p_j) (p_i - p_i^*) \right]$$
$$= \sum_{i,j} A_{ij} (p_j^* - p_j) (p_i - p_i^*)$$

\uparrow used $r_i = \sum_j A_{ij} p_j^*$

Now use the form $A_{ij} = \sum_\alpha V_{i\alpha} V_{j\alpha} \sigma_\alpha$

$$\begin{aligned} \frac{dL}{dt} &= - \sum_\alpha \sigma_\alpha \sum_{i,j} V_{i\alpha} V_{j\alpha} (p_i - p_i^*) (p_j - p_j^*) \\ &= - \sum_\alpha \sigma_\alpha \left(\sum_i V_{i\alpha} (p_i - p_i^*) \right)^2 \leq 0 \end{aligned}$$

$\Rightarrow L(t)$ always decreasing until $p_i(t) = p_i^*$

where $L(t) = 0$. i.e. global attractor!

\Rightarrow Only pending on the existence of $p_i^* > 0$

c) Solve for fixed point (feasibility)

- Start with $r_i = \sum_j A_{ij} p_j$

$$\begin{aligned} \tilde{r}_{j\alpha} &= \sigma_\alpha^{1/2} v_{j\alpha} \\ &= \left(\frac{K_\alpha}{\gamma_\alpha Y_\alpha} \right)^{1/2} v_{j\alpha} \end{aligned}$$

$$\begin{aligned} &= \sum_{j\alpha} \sigma_\alpha v_{j\alpha} \tilde{r}_{j\alpha} p_j \\ &= \sum_{j\alpha} \tilde{v}_{j\alpha} \tilde{r}_{j\alpha} p_j \end{aligned}$$

Matrix notation: $A = \tilde{v} \times \tilde{v}^T$
where $(\tilde{v}^T)_{\alpha i} = \tilde{v}_{i\alpha}$

$\tilde{v}_{i\alpha}$: Consumption rate of resource α by consumer i
 N_R (# col.) N_C (# row)

$A = \tilde{v} \times \tilde{v}^T$: Symmetric square matrix
 $(N_C \times N_C)$

Linear Algebra:

If matrix \tilde{v} is full rank with $N_C \leq N_R$

then $A = \tilde{v} \times \tilde{v}^T$ is invertible

$$\Rightarrow p_i^* = \sum_j A_{ij}^{-1} r_j$$

where $A^{-1} = (\tilde{v} \times \tilde{v}^T)^{-1}$, $r_i = \sum_\alpha v_{i\alpha} K_\alpha - M_i$

Special case: $N_R = N_C$, $p_i^* = \sum_\alpha \tilde{v}_{i\alpha}^T Y_\alpha Y_\alpha - O(\mu)$

cf. The random matrix perspective:

(connect to May's work — Cui et al., 2019)

Effective SLE system:

$$\frac{d}{dt} \rho_i = r_i \rho_i - \sum_{j=1}^{N_c} A_{ij} \rho_i \rho_j$$

$$\text{where } A_{ij} = \sum_{\alpha} v_{i\alpha} v_{j\alpha} \sigma_{\alpha}$$

perturbation from fixed pt $\rho_i^* = \sum_j A_{ij}^{-1} r_j$

$$\delta \rho_i = \rho_i - \rho_i^*$$

$$\frac{d}{dt} \delta \rho_i = - \rho_i^* \sum_j A_{ij} \delta \rho_j$$

$$\text{Community matrix } M_{ij} = -\rho_i^* A_{ij}$$

R. May: $M_{ij} = R_{ij} - \delta_{ij}$ if $\text{Var} = \sigma^2$

↑ random non-neg elements

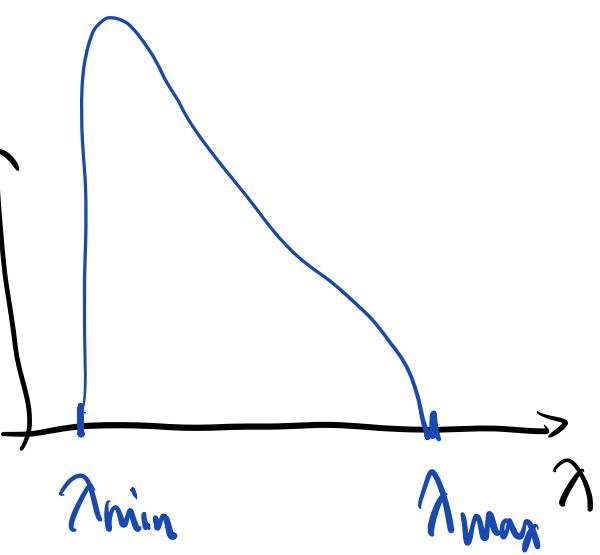
→ largest eigenvalue $= -1 + \sigma \sqrt{N_c} > 0$ for $N_c \gg 1$.

For the CR model, $A = v \cdot v^T$

$v_{i\alpha} = \text{iid, non neg.} (\text{Var} = \sigma_v^2) \rightarrow \text{Wishart matrix}$

eigenvalue dist (Marchenko-Pastur dist — for Gaussian dist of $v_{i\alpha}$)

$$P(\lambda) = \frac{Nr/N_c}{2\pi\sigma_v^2 \lambda} \sqrt{(\lambda_{\max} - \lambda)(\lambda - \lambda_{\min})}$$



$$\lambda_{\min} = \sigma_v^2 \left(1 - \sqrt{N_c/N_R}\right)^2$$

$$\lambda_{\max} = \sigma_v^2 \left(1 + \sqrt{N_c/N_R}\right)^2.$$

→ $\lambda \geq 0$ as long as $N_c \leq N_R$
Even as $N_c, N_R \rightarrow \infty$.

Relation to community matrix M :

largest eigenvalue of M_{ij}

= - smallest eigenvalue of A_{ij}^\top = $-\lambda_{\min}$

→ Community matrix is stable as $N_c, N_R \rightarrow \infty$,
as long as $\frac{N_c}{N_R} < 1$.

→ In practice, even if $N_c = N_R$,
feasible soln ($P_i^* > 0$) involves $N_c^* < N_R$.

* Recent Study (Cui et al, 2019) showed numerically that for large random consumption matrix $V_{1,2}$, typically $\sim 50\%$ of $V_{1,2}$ exhibit coexistence. (no analytical result on this so far)

