## Supplementary Material for

"Diversity begets stability: sublinear growth scaling and competitive coexistence across ecosystems"

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# A Conditions for stable coexistence in competitive communities

Here we provide more details on the analytical results discussed in the main text.

#### A.1 Homogeneous communities: stability criteria

Let us model the competition of S species with abundances  $x_i$  (either numerical density  $N_i$  or biomass density  $B_i$ ) in terms of per capita growth rates, as follows

$$\frac{1}{x_i}\frac{dx_i}{dt} = \text{inter}_i(x_i) + \text{intra}_i(\mathbf{x}_{-i}) \equiv f_i(\mathbf{x}), \quad 1 \le i \le S.$$
(A.1)

Here inter<sub>i</sub> represents intra-specific effects which only depend on the abundance of population i, and intra<sub>i</sub> represents inter-specific effects which only depend on the abundances of its competitors, collectively denoted  $\mathbf{x}_{-i} \equiv (x_j)_{j\neq i}$ . Because competition can only reduce the growth rate of a population, we assume that both terms are monotonically decreasing functions of their arguments.

Following standard terminology, the system may be called *feasible* if it admits an equilibrium point  $\mathbf{x}^*$  whose components  $x_i$  are all positive. Further, an equilibrium point  $\mathbf{x}^*$  is (locally) stable if all eigenvalues of the Jacobian of f evaluated at  $\mathbf{x}^*$  (also known as the 'community matrix'  $C^*$  of the system) have negative real parts. The community matrix has diagonal and off-diagonal elements, respectively

$$C_{ii}^* = x_i^* \frac{\partial \text{intra}_i}{\partial x_i}(x_i^*), \quad C_{ij}^* = x_i^* \frac{\partial \text{inter}_i}{\partial x_j}(\mathbf{x}_{-i}^*).$$
 (A.2)

Consider the case species are indistinguishable and interactions are uniform  $(A_{ij} = A > 0 \text{ for } i \neq j)$ . In that case all species have equal abundance  $x^*$  at equilibrium, and moreover the eigenvalue  $\lambda$  of the Jacobian  $J^*$  with largest real part is given by

$$\lambda = x^* \left( \frac{\partial \text{intra}}{\partial x_i} (x^*) - \frac{\partial \text{inter}}{\partial x_j} (\mathbf{x}^*) \right). \tag{A.3}$$

From this expression derives a general rule of thumb concerning the stability of competitive communities: a multi-species equilibrium can only be stable if *intra-specific regulation is more strongly density-dependent than inter-specific competition*.

For the model introduced in the main text, we have  $\operatorname{intra}(x_i) = rg(x_i) - z$  and  $\operatorname{inter}(x_{-i}) = -\sum_{j \neq i} A_{ij} x_j$ . We can therefore write the local stability condition as

$$rg'(x^*) < -A. (A.4)$$

In the logistic case we have  $g'(x^*) = -1/K$ , hence stability requires A < r/K: interspecific interactions must be weaker than intra-specific interactions. In the sublinear model, on the other hand, we have  $g'(x^*) = (k-1)g(x^*)/x^*$ , and using the equilibrium condition  $rg(x^*) - z = (S-1)Ax^*$ , we arrive at  $(1-k)[(S-1)A + z/x^*] > A$ . This condition is fulfilled whenever

$$S > 1 + \frac{1}{1 - k}. (A.5)$$

(When z > 0 this condition is sufficient but not necessary for local stability.) Eq. (A.5) shows that, at least in homogeneous communities, increasing diversity enhances the local stability of multi-species equilibria.

### A.2 Homogeneous communities: May vs. anti-May behavior

In this section we show that the GLV model lies on the boundary between two phases: one with May behavior (diversity begets instability), and one with anti-May behavior (diversity begets stability). To this aim we consider a variant of the competition model of the form

$$\frac{1}{x_i}\frac{dx_i}{dt} = x_i^k - \sum_{j=1}^S a_{ij}x_i^a x_j^b = x_i^k - A_{ii}x_i^{a+b} - \sum_{j\neq i} a_{ij}x_i^a x_j^b.$$
 (A.6)

In this formulation the GLV model corresponds to k = a = b = 1 and the sublinear model to k < 1 and a = b = 1. We compute the Jacobian matrix elements:

$$C_{ii} = kx_i^{k-1} - A_{ii}(a+b)x^{a+b-1} - a\sum_{j\neq i} A_{ij}x_i^{a-1}x_j^b,$$
(A.7)

$$C_{ij} = -b \sum_{j \neq i} A_{ij} x_i^a x_j^{b-1}. (A.8)$$

If we assume as before a homogeneous community with identical interaction strengths (denoted  $A_{ii} = A_s$ ,  $A_{ij} = A_c$ ), the equilibrium abundance reads

$$x^* = [A_s + (S-1)A_c]^{1/(k-a-b)}$$
(A.9)

and the corresponding community matrix

$$C_{ii} = k(x^*)^{k-1} - [(a+b)A_s + a(S-1)A_c](x^*)^{a+b-1},$$
  
 $C_{ij} = -bA(x^*)^{a+b-1} \quad \text{for } i \neq j.$ 

The largest eigenvalue of  $C^*$  is given by

$$\lambda = C_{ii}^* - C_{ij}^*$$

$$= k(x^*)^{k-1} - [(a+b)A_s + (a(S-1)-b)A_c](x^*)^{a+b-1}$$

$$= \left(k - \frac{(a+b)A_s + (a(S-1)-b)A_c}{A_s + (S-1)A_c}\right)(x^*)^{k-1}.$$

The local stability condition is, therefore.

$$(S-1)(k-a) < (a+b-k)\alpha - b$$
 with  $\alpha \equiv A_s/A_c$ . (A.10)

We have three possibilities:

$$S > 1 + \frac{(a+b-k)\alpha - b}{k-a}$$
 if  $k < a$ ,  

$$S < 1 + \frac{(a+b-k)\alpha - b}{k-a}$$
 if  $k > a$ ,  

$$\alpha > 1$$
 if  $k = a$ .

A series of comments are in order:

- Depending on k and a we have two regimes: one in which increasing S hinders stability (k > a, a 'May' phase), and one in which increasing S enhances stability (k < a, an 'anti-May' phase). Notably this transition is independent from b, indicating that only the relative density dependence of self- and cross-regulation in the *per capita* growth of a population is relevant, and not the form of the dependence on other species' abundance in this *per capita* growth.
- The critical case (k = a) recovers the usual stability condition for the homogeneous GLV model (b = 1) and generalizes it for general interactions  $(b \neq 1)$ .
- For a = b = 1 and  $A_s = 0$  we recover the result (A.5) obtained in the previous section.

These results are obtained in the special case where interactions are homogenous (standard deviation  $\sigma = 0$ ). When  $\sigma > 0$ , the GLV model falls squarely into the May phase, as illustrated e.g. by Fig. 3A of the main text.

#### A.3 Random communities: DMFT and cavity solution

Dynamical mean field theory (DMFT) and the cavity method have been used to study the GLV model in the large diversity limit  $(S \to \infty)$  by several authors [REFS]. Here we show how to derive the derive analytical results in the sublinear model using these techniques.

#### A.3.1 Equilibrium distribution

#### A.3.2 Stability condition

Ahmadian et al. [AHMADIAN] consider large matrices of the form M + LJR, where M, L and R are deterministic matrices, and J is a random matrix with i.i.d. coefficients with zero mean and variance  $\sigma^2$ . Generalizing the classical 'circular law', they show that the spectrum of such matrices is contained withing the region of the complex plane defined by

$$\text{Tr}[(M_{\zeta}M_{\zeta}^{\dagger})^{-1}] \ge 1/\sigma^2 \quad \text{where } M_{\zeta} = L^{-1}(\zeta I - M)R^{-1},$$
 (A.11)

where Tr denotes the trace, dagger the Hermitian conjugate, and  $\zeta \in \mathbb{C}$ . In the special case where L, R and M are diagonal, this condition reduces to

$$\sum_{i=1}^{S} \frac{(L_i R_i)^2}{|\zeta - M_i|^2} \ge 1/\sigma^2 \tag{A.12}$$

for matrices of size S.

Following Stone [STONE], we now apply this result to generalize the stability condition to random interactions  $A_{ij}$ . From (A.2) we can write the community matrix as

$$C^* = -\operatorname{diag}(\mathbf{x}^*)[A + \operatorname{diag}(\mathbf{r}g'(\mathbf{x}^*))] \tag{A.13}$$

where the product of vectors is understood component-wise. Denoting 1 the matrix with all entries equal to 1, we can write  $A = \mu \mathbf{1} - \mu I + J$  with J as above, hence  $C^*$  has the form M + LJR with  $L = -\text{diag}(\mathbf{x}^*)$ , R = I and  $M = \text{diag}(\mathbf{x}^*)[\mu I - \text{diag}(\mathbf{r}g'(\mathbf{x}^*))]$ . Thus, we have that the eigenvalues of the community matrix  $C^*$  must lie within the domain

$$\sum_{i=1}^{S} \frac{(x_i^*)^2}{|z - x_i^*[\mu - r_i g'(x_i^*)]|^2} \ge 1/\sigma^2.$$
(A.14)

This domain first touches the right half-plane at z=0, hence stability of  $C^*$  requires

$$\sum_{i=1}^{S} \left[ \mu - r_i g'(x_i^*) \right]^{-2} < \sigma^{-2}. \tag{A.15}$$

In the logistic model, we have  $g'(x_i^*) = -1/K_i$ , hence (A.15) becomes  $\sum_i (\mu - r_i/K_i)^{-2} < \sigma^{-2}$ . When growth rates and carrying capacities are all equal, this is just the May condition  $\sigma \sqrt{S} + \mu < r/K$ , interpreted as stating that interaction strength

 $(\mu, \sigma)$  and diversity S must be small compared to self-regulation r/K to allow for stable coexistence.

By contrast, in the sublinear model (defined by  $g(x) = (x/x_0)^{k-1}$ ), we have  $g'(x) = (k-1)x^{k-2}/x_0^{k-1}$ , hence (A.15) reads

$$\sum_{i=1}^{S} \left[ \mu + (1-k)r_i(x_i^*)^{k-2}/x_0^{k-1} \right]^{-2} < \sigma^{-2}.$$
(A.16)

Making use of the cavity solution  $P(x^*)$  obtained in sec. A.3, we can write this condition as

$$S \int dP(r, x^*) [\mu + (1 - k)r(x^*)^{k-2}/x_0^{k-1}]^{-2} < \sigma^{-2}$$
(A.17)

TBC

### A.4 Random communities: replica solution

# B Linking dynamical theory with macroecological laws

This section expands on the way we parametrize the sublinear model using individual-level allometries to recover various macroecological patterns.

- **B.1** Production-biomass scaling
- B.2 Species abundance distribution
- B.3 Mean-variance scaling
- B.4 Size-density scaling
- B.5 Size spectra