

Omnomnivores

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7 **Abstract**
8 TODO

9 **1 Introduction**
10 **2 Data & Methods**
11 **2.1 Metacommunity model**

12 The model used broadly follows the metacommunity model developed by Thompson
13 & Gonzalez (2017). The model (Equation 1) itself is based on a tritrophic com-
14 munity (‘plants’, ‘herbivores’, and ‘carnivores’), and is essentially a collection of
15 modified Lotka–Volterra equations, this (broadly) models species abundance as a
16 function of interaction strength, environmental effect, immigration, and emigration.
17 The metacommunity consists of S species with M environmental patches in the
18 landscape and looks as follows:

$$X_{ij}(t+1) = X_{ij}(t) \exp \left[C_i + \sum_{k=1}^S B_{ik} X_{kj}(t) + A_{ij}(t) \right] + I_{ij}(t) - X_{ij}(t) a_i \quad (1)$$

19 Where $X_{ij}(t)$ is the abundance of species i in patch j at time t . C_i is its intrinsic
20 rate of increase (which we have set to 0.1 for ‘plants’ and -0.001 for ‘herbivores’
21 and ‘carnivores’). B_{ik} is the per capita effect of species k on species i . The exact
22 interaction strength for each species pair is determined by the trophic level of each
23 species and is drawn from a uniform distribution. The ranges for each combination
24 of species pairs listed in Table 1, the values that are drawn from the uniform distri-
25 bution are then scaled by dividing by 0.33 S to yield the final interaction strength for
26 each interacting pair.

Table 1: Intervals used for the uniform distribution from which interaction strengths values are drawn from for the different types of species pair interactions. Note this is represent the effect of species type 1 on species type 2 *i.e.*, herbivore-plant represents the effect of a herbivore species on a plant species

Interacting pair	Range of uniform distribution
Plant-plant	-1.0 – 0.00
Plant-herbivore	0.0 – 0.10
Plant-carnivore	0.0
Herbivore-plant	-0.3 – 0.00
Herbivore-herbivore	-0.2 – -0.15
Herbivore-carnivore	0.0 – 0.08
Carnivore-plant	0.0
Carnivore-herbivore	-0.1 – 0.00
Carnivore-carnivore	-0.1 – 0.00

27 $A_{ij}(t)$ is the effect of the environment in patch j on species i at time t . Essentially
28 this will set the effect of the environment to zero when it is at the optimum of the
29 species and can be further expanded as follows:

$$A_{ij}(t) = \left(h \times \frac{1}{\sigma\sqrt{2\pi}} \right) \times \left(\exp \left[-\frac{(E_j(t) - H_i)^2}{2\sigma^2} \right] - 1 \right) \quad (2)$$

30 Where the species environmental optima (H_i) are evenly distributed across the entire
31 range of environmental conditions for each trophic level, meaning that species

32 from different trophic levels will be at, or near the same environmental optima. h is
 33 a scaling parameter (set to **50**), $E_j(t)$ is the environment in patch j at time t and σ
 34 is the standard deviation (set to **50**).

35 $I_{ij}(t)$ is the abundance of species i immigrating to patch j at time t and can be
 36 expanded as follows:

$$I_{ij}(t) = \sum_{l=j}^M a_i X_{il}(t) \exp(-Ld_{jl}) \quad (3)$$

37 Where a_i is the proportion of the population of species i that disperses at each time
 38 step, the dispersal rate is drawn from a normal distribution ($\mu = 0.1$, $\sigma = 0.025$) for
 39 each species. The abundance of immigrants to patch j from all other patches is gov-
 40 erned by where d_{jl} is the geographic distance between patches j and l , and L (the
 41 strength of the exponential decrease in dispersal with distance), which is also drawn
 42 from a normal distribution for each species. The parameters used for L are trophic
 43 level dependant and are show in Table 2

Table 2: Parameters for the normal distributions used to determine the dispersal decay (L) for each species depending on its trophic level.

Trophic level	μ	σ
Plant	0.3	0.075
Herbivore	0.2	0.050
Carnivore	0.1	0.025

44 2.2 Generating networks

45 In order to create a final community state the species are allowed to persist for a
 46 total of 2000 generations. These generations are broken down into three ‘phases’
 47 the first is the ‘proofing’ phase where the environment is uniform throughout the
 48 landscape (meaning that all species are at their environmental optimum) for 500
 49 generations. After this the environment is ‘heated’ incrementally until it reaches its
 50 ‘final state’, the environmental optimum of each species is also adjusted as the en-
 51 vironmental values begin to change. This occurs over a period of 1 000 generations.
 52 The landscape is then held stable for a further 500 generations until an equilibrium
 53 is reached. The final state of the landscape is predetermined and is defined by the
 54 diamond-square algorithm (this produces fractals with variable spatial autocorre-
 55 lation) which is generated using `NeutralLandscapes.jl` (Catchen, 2023), here we
 56 vary the degree of landscape heterogeneity by **TODO**.

Table 3: Starting parameters for the model.

Parameter	Value
S	100
M	26*26
$E_{initial}$	40
$A_{initial}$	0.01

57 2.3 Spatial wombling

58 Broadly speaking spatial wombling is an edge-detection algorithm which traverses
 59 a geographic area and defines this area in terms of the rate (m) and corresponding

60 direction (θ) of change. This is done by using first-order partial derivative (∂) of the
61 ‘curvature’ of the landscape as described by $f(x, y)$ (see Equation 4). This essentially
62 gives an indication how steep the gradient (m) is between neighbouring cells as well
63 as the direction (θ) of the slope.

$$m = \sqrt{\frac{\partial f(x, y)^2}{\partial x} + \frac{\partial f(x, y)^2}{\partial y}} \quad (4)$$

64 The spatial wombling analyses were done using `SpatialBoundaries.jl` (Strydom
65 & Poisot, 2023). The documentation provides a more detailed breakdown of the
66 underlying methodology.

67 **3 Conclusion**

68 **References**

69 Source: [Article Notebook](#)

70 Catchen, M. D. (2023, December). `EcoJulia/NeutralLandscapes.jl`. EcoJulia.

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