

Neutral competition boosts chaos in food webs

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Abstract

14 Near-neutrality of competition has been proposed to facilitate coexis-
tence of species because it slows down competitive exclusion, thus mak-
16 ing it easier for equalizing mechanisms to maintain diverse communities.
An unrelated line of work has shown that chaos can promote coexis-
18 tence of many species in super-saturated communities. By analyzing a
set of numerically simulated food webs, here we link those previously
20 unrelated findings. We show that near-neutrality of competition at the
prey's trophic level, in the presence of interactions with natural enemies,
22 increases the chances of developing chaotic dynamics. Our results sug-
gest that near-neutrality may promote biodiversity in two ways: through
24 reducing the rates of competitive displacement and through promoting
non-equilibrium dynamics.

1 Background

26

Ever since Darwin, the idea that species must be sufficiently different to be able to coexist is deeply rooted in the history of biological thinking. Indeed, the principle of competitive exclusion is intuitively straightforward, and elegant mathematical underpinning (Macarthur and Levins, 1967) helped making it one of the cornerstones of ecological theory. Nevertheless, on a closer examination, natural communities often seem to harbor far more species that may be reasonably explained from niche separation. Plankton communities, where many species coexist with little room for differentiation, have served as an early example (Hutchinson, 1959, 1961), inspiring the legendary ecologist G. Evelyn Hutchinson to ask the simple but fundamental question "*why are there so many kinds of animals?*" (Hutchinson, 1959). Since then many mechanisms have been suggested that may help similar species to coexist. As Hutchinson already proposed himself, fluctuations in conditions may prevent reaching equilibrium at which species would be outcompeted. Also, natural enemies including pests and parasites tend to attack the abundant species more than rare species, and such a "*kill the winner*" (Winter et al., 2010) mechanism promotes diversity by preventing one species to take all the resources and outcompete the rest.

In the extensive literature on potential mechanisms that could prevent competitive exclusion there are two relatively new ideas that radically differ from the rest and have created some controversy: neutrality and chaos. The neutral theory of biodiversity introduced by Hubbell (Hubbell, 1997) proposes that species that are entirely equivalent can coexist in a neutral way because none is able to outcompete the other. The concept of completely equivalent species has met skepticism as it is incompatible with the idea that all species are different. However, it turns out that also "*near-neutrality*" arises robustly in models of competition and evolution and may boost the chances for coexistence (Scheffer and van Nes, 2006; Scheffer et al., 2018; Fort et al., 2009, 2010). Support for such near-neutrality has been found in a wide range of communities (Scheffer

and van Nes, 2006; Vergnon et al., 2013; Scheffer et al., 2015; Segura et al.,
 56 2013; Vergnon et al., 2012). The second relatively new and controversial mech-
 anism that may prevent competitive exclusion is *"super-saturated coexistence"*
 58 in communities that display chaotic dynamics (Huisman and Weissing, 1999).
 This is in a sense analogous to the prevention of competitive exclusion in fluc-
 60 tuating environments, except that deterministic chaos may arise in autonomous
 non-linear systems without any external fluctuation. Although there has been
 62 much debate about the question whether such internally driven complex dy-
 namics plays an important role in ecosystems, several studies support the idea
 64 that chaos can be an essential ingredient of natural dynamics (Armstrong and
 McGehee, 1980; Huisman and Weissing, 1999; Benincà et al., 2008; Dakos et al.,
 66 2009).

Intuitively, ~~a relation between chaos and neutrality seems unlikely~~, as fully
 68 neutral ecosystems can't be chaotic. But natural ecosystems are never perfectly
 neutral. In the present work, we used a multi-species food-web model to explore
 70 the effect of near-neutrality at the prey level on the probability of developing
 chaotic dynamics. We found a surprising relationship between both ideas: the
 72 closer to neutrality the competition is, the higher the chances of developing
 chaotic dynamics.

74 2 Methods

2.1 Model description

76 We focused our attention on food webs with two trophic levels, one of con-
 sumers and another of prey. The consumers predate on the prey, and the prey
 78 populations are subject to implicit competition.

The dynamics were modelled using the Rosenzweig-MacArthur predator-
 80 prey model (Rosenzweig and MacArthur, 1963), generalized to a higher number
 of species (see van Nes and Scheffer (2004)). Our model contains n_P prey species

and n_C predator species. The prey's populations are under the influence of both intra and interspecific competition, whose intensities are coded in the competition matrix A . The relative preference that predators show to each prey is coded in the predation matrix S . In the absence of prey, the predators' populations just decay exponentially. Prey immigration from neighboring areas has been added to the classical model in order to avoid unrealistic dynamics, such as heteroclinic orbits giving rise to long-stretched cycles with near extinctions (van Nes and Scheffer, 2004). In mathematical notation, the system reads:

$$\begin{cases} P'_i(t) = r_i(P)P_i - \sum_{j=1}^{n_C} g_j(P)P_i S_{ji}C_j + f & : i = 1..n_P \\ C'_j(t) = -lC_j + e \sum_{i=1}^{n_P} g_j(P)P_i S_{ji}C_j & : j = 1..n_C \end{cases} \quad (1)$$

where $P_i(t)$ represents the biomass of prey species i at time t and $C_j(t)$ the biomass of predator species j at time t . The symbol P is used as a shorthand for the vector $(P_1(t), P_2(t), \dots, P_{n_P}(t))$. The auxiliary functions $r_i(P)$ and $g_j(P)$ (see equations (2) and (3)) have been respectively chosen to generalize the logistic growth and the Holling type II saturation functional response (Edelstein-Keshet, 2005) to a multispecies system when inserted into equation (1).

$$r_i(P) = r \left(1 - \frac{1}{K} \sum_{k=1}^{n_P} A_{ik}P_k \right) \quad (2)$$

$$g_j(P) = \frac{g}{\sum_{i=1}^{n_P} S_{ji}P_i + H} \quad (3)$$

For details about the parameters used, please refer to subsection 2.2.

2.2 Parameterization

We parameterized our model as a freshwater plankton system based on Dakos' model (Dakos et al., 2009). Dakos' model uses a Rosenzweig-McArthur multispecies model with two trophic levels, and parameterizes it to describe a zooplankton - phytoplankton system. Unlike Dakos, who uses seasonally changing parameters, our parameters were assumed to be independent of time (see table 1).

| Symbol | Interpretation | Value | Units |
|--------|---|-------------------|----------------------|
| r | Maximum growth rate | 0.50 | d^{-1} |
| K | Carrying capacity | 10.00 | $mg\ l^{-1}$ |
| g | Predation rate | 0.40 | d^{-1} |
| f | Immigration rate | 10^{-5} | $mg\ l^{-1}\ d^{-1}$ |
| e | Assimilation efficiency | 0.60 | 1 |
| H | Saturation constant | 2.00 | $mg\ l^{-1}$ |
| l | Predator's loss rate | 0.15 | d^{-1} |
| S | $n_C \times n_P$ predator preference matrix | See section 2.2.1 | 1 |
| A | $n_P \times n_P$ competition matrix | See section 2.2.1 | 1 |

Table 1: Values and meanings of the parameters used in our numerical experiment. The elements of the predation (S) and competition (A) matrices are drawn from probability distributions described in subsection 2.2.1.

2.2.1 Competition and predation matrices

Our main purpose is to analyze the effect of different competition strengths on the long term dynamics exhibited. In order to simulate and quantify this heterogeneity, we introduce the competition parameter ϵ . This dimensionless parameter was used to build a competition matrix A , whose diagonal terms are identically 1, and whose non-diagonal terms are drawn from a uniform probability distribution centered at $1 + \epsilon$ and with a given width (here we chose $w = 0.05$). Defined this way, the parameter ϵ allows us to travel continuously from strong intraspecific ($\epsilon < 0$) to strong interspecific competition ($\epsilon > 0$), meeting neutral competition near $\epsilon = 0$. Two ecosystems with the same competition parameter, despite not being identical, will exhibit the same competition type. For the rest of this paper, we will call *neutral-on-average* those ecosystems with $\epsilon = 0$ (see figure 1).

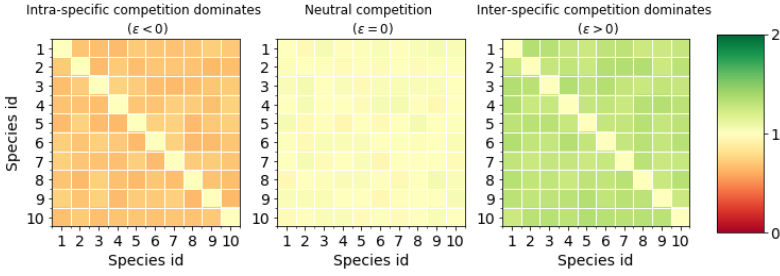


Figure 1: The competition matrix on the left is a clear case of dominant intra-specific competition. The central one represents a case of neutral-on-average competition. The matrix in the right panel shows a case of dominant inter-specific competition. The difference between them is the relative size of the non-diagonal elements respective of the diagonal ones. This property of the competition matrices is controlled by the competition parameter ϵ .

Regarding the predation matrix S , we proceed as in Dakos et al. (2009) and draw each of its coefficients from a uniform probability distribution bounded between 0 and 1.

2.3 Numerical experiment

Depending on the parameters and initial conditions, a system like the one described in equation (1) can give rise to three types of qualitative behaviour, each of them roughly corresponding to a different type of attractor (see figure 2). The first one, a stable point attractor, generates a constant species composition. Secondly, limit cycle (and limit tori) attractors give rise to periodically (or quasiperiodically) changing species composition. Lastly, we'll refer as chaotic to attractors that, while remaining bounded, do not fit in any of the previous categories.

Our target is to estimate the probability of reaching each type of attractor under different assumptions about competition. In order to achieve this, we swept among 25 different values of the competition parameter ϵ (defined in

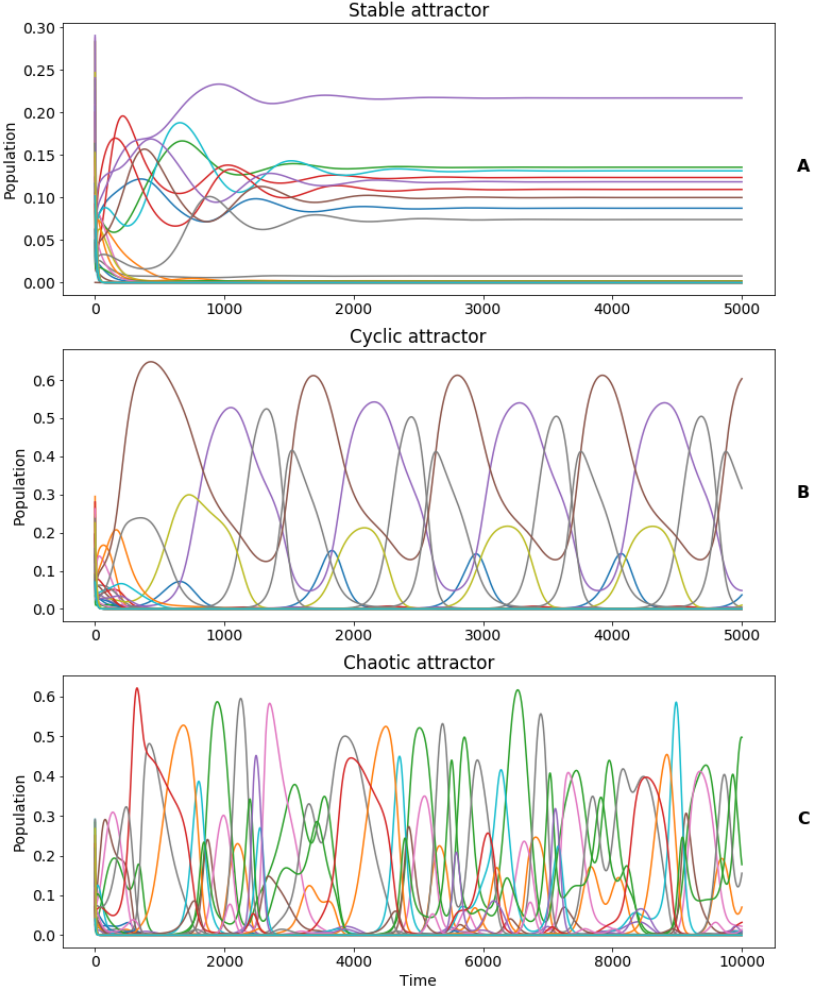


Figure 2: Our family of models generates time series of the population of each species. The time series can be classified in 3 qualitative types depending on their asymptotic behaviour: *stable*, *cyclic* and *chaotic*. In **panel A**, the system reaches a stable attractor after a transient time. In **panel B**, a periodic attractor, with an approximate period of 1000 days, is reached after the transient time. The system in **panel C** never reaches a stable nor a cyclic attractor, but a chaotic one.

section 2.2.1), from $\epsilon = -0.8$ to $\epsilon = 0.8$. The lower value was chosen to assure that the non-diagonal competition matrix elements can never be very close to zero, so our network remains always ~~highly~~ connected. The upper value was motivated by symmetry with the lower one. For each of the 25 values of our competition parameter, 200 different predation and competition matrices were drawn from the probability distributions described in section 2.2.1. Together with equation (1) each of those parameter sets represents an ecosystem. We used standard numerical methods to solve each realization of equation (1) and, thus, simulating the corresponding ecosystem. A first stabilizing run of 2000 days was executed in order to get closer to the attractor. Simulating for 5000 more days, we obtained time series in the vicinity of this attractor.

We used the Gottwald - Melbourne test (see Gottwald and Melbourne (2009)) to classify each of our time series as *stable*, *cyclic* or *chaotic*. ~~The ratio of attractors found to be chaotic can be used to estimate the probability of a given family of ecosystems to develop chaotic behaviour.~~ The probabilities of stable or cyclic behaviour can be estimated analogously. Additionally, two different measures of biodiversity were applied to each simulated ecosystem: the average number of non-extinct prey species and the average biomass grouped by trophic level. We considered ~~extinct those species whose~~ population density remained below a threshold of 0.01 mg l^{-1} after the stabilization run. ~~Grouping our simulated ecosystems by competition parameter ϵ , we explored~~ the relationship between the competition strength, the probability of each dynamical regime and the biodiversity.

The numerical experiment was repeated for food webs of different sizes, ranging from 5 to 50 species. In our simulations, we kept a ratio of 2:3 for the number of species at the consumer and the prey level.

In the spirit of reproducible research, we made available the code used to obtain our conclusions and generate our figures (Rodríguez-Sánchez, 2018).

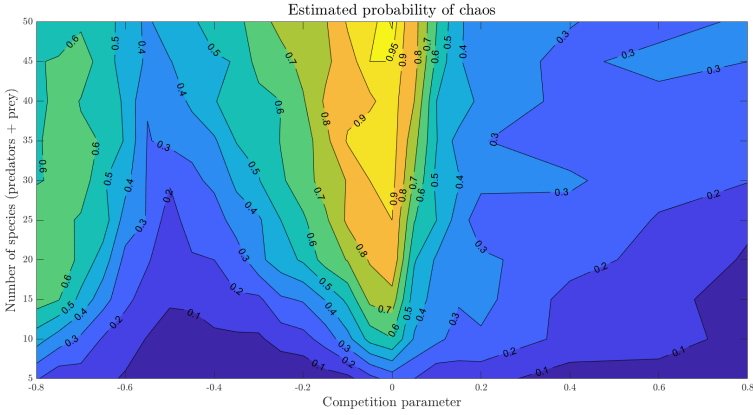


Figure 3: Contour map showing the probability of chaos for various competition parameters (horizontal axis) and number of species (vertical axis). The consumers’ population is fixed as $2/3$ of the prey’s population. Notice that chaotic attractors appear more easily (i.e., for smaller systems) the closer is the competition to neutral (i.e., $\epsilon = 0$).

3 Results

From figure 3 (see also figure A.1 in the Online Appendix) we conclude that, in our model, the likelihood of chaotic behaviour for neutral-on-average competition at the prey level is higher than for dominant inter or intraspecific competition. This result remains true for systems with a different number of species. The likelihood of chaos also increases with the size of the food web. This effect should not be surprising: the more dimensions the phase space has, the easier is to fulfill the requirements of the complex geometry of a chaotic attractor (Strogatz, 1994). Even in those higher dimensional cases, there is still a clear maximum at neutral-on-average competition. The probability of chaos shows another local, lower maximum for weak competition coupling. We consider this a reasonable result, as predation is known to be the main driver of chaos in this kind of models (van Nes and Scheffer, 2004).

Our biodiversity measurements (see figure 4) illustrate several additional

effects. In panel C we see that the prey biodiversity is very close to its maximum around the neutral on-average situation. Also, a maximum is observed for strong intraspecific competition. In panels B and C, we see that the dynamic regime has an influence in the prey biodiversity. Particularly, the average prey biodiversity for ecosystems with stable dynamics is always lower than for cyclic ones, and both of them are lower than for those with chaotic dynamics. Despite the dispersion of our box and whisker plot is high, this conclusion remains true for food webs of different sizes (see figure A.4 in the Online Appendix). From panel D, we see that the prey biomass remains relatively stable for the whole range of competition parameters, with the exception of strong intraspecific competition, where it reaches a maximum. The predator biomass grows almost linearly as the competition moves leftwards, from neutral-on-average to strong intraspecific, while the prey biomass remains constant. This effect is known from 2 species systems with Type II saturation functions.

4 Discussion

The asymptotic dynamics of our model are affected by the strength of intraspecific competition compared to interspecific competition. Interestingly, we find that competition close to neutrality significantly increases the chances of chaotic behaviour. This suggests that in a system with predation, neutrality-on-average at the competition level may increase the probability of complex dynamics if the species are not equally prone to predation. We also found a clear correlation between non-stable dynamics and a higher prey biodiversity, being chaotic dynamics more efficient than cyclic at harboring a higher biodiversity. This allows us to conclude that chaos improves the chances for a higher number of coexisting species in our model. These observations, when put together, suggest that the hypothesis of non-equilibrium (Huisman and Weissing, 1999) and Hubbell's hypothesis of neutrality are not completely independent (see figure 5).

In the spirit of mathematical modelling, we chose the simplest realization

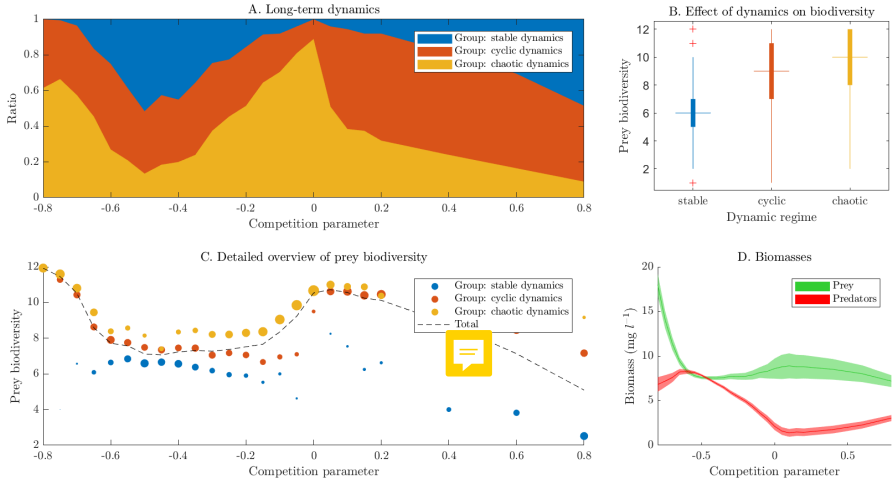


Figure 4: Results for a food web with 8 predator and 12 prey species. Food webs of different sizes show similar results (see A.1 in Online Appendix). For each value of the competition parameter, 200 randomly drawn ecosystems were simulated and classified as regular, cyclic or chaotic. Additionally, the number of non-extinct prey species after a stabilization run was registered. **Panel A.** Ratio of each dynamic regime vs. competition parameter. **Panel B.** Biodiversity vs. asymptotic regime. Box and whisker plot of the average number of non-extinct prey species grouped by asymptotic regime. **Panel C.** Average prey biodiversity vs. competition parameter. The dashed line shows the average number of non-extinct prey species grouped by competition parameter. The colored circles represent the average prey biodiversity of the simulations, additionally grouped by dynamical regime (stable, cyclic and chaotic). The relative size of the circles represents the ratio of simulations that led to regular or chaotic dynamics. **Panel D.** Average biomasses grouped by trophic level vs. competition parameter. The width represents standard deviation.

required for the effects to be observed. We explored the variation away from
neutrality only along the stabilizing/destabilizing axis. We didn't use Allee
effect, nor noise, nor species-specific carrying capacities, and the functional form
of each term has been chosen to account for satiation and saturation in the
simplest possible ways. This opens the door to perform similar analyses in the
future using more sophisticated models.

Both the competition and predation parameter sets were drawn from prob-
ability distributions. The interactions in our system can be interpreted as a
weighted network with a high connectivity. In nature, trophic networks tend to
show modular structure with various clusters (Thebault and Fontaine, 2010).
Our simplified model could be interpreted as representing one of those densely
connected modules. It is known that the asymptotic behaviour of this kind
of systems can be very sensitive to the parameters choice. In particular, in-
troducing correlations between parameters can greatly modify the probabilities
of chaotic attractors to be reached (see for instance Huisman et al. (2001), in
response to the letter Schippers et al. (2001)). In the present paper we didn't
introduce any correlation, i.e., all our random parameters were drawn indepen-
dently from the others. Studying the effect of different physiological scenarios
(in the sense of Huisman et al. (2001), that is, constraints between the param-
eters) on the probabilities of chaos could be a continuation to this paper.

Due to the large number of simulations made, we had to rely on automatic
methods for detecting chaos. Automatic detection of chaos by numerical meth-
ods has fundamental limitations, especially for high dimensional systems like
ours. Most of them can be boiled down to the fact that, in general, numeri-
cal methods cannot distinguish robustly between long, complicated transients
and genuine chaos. Our motivation to chose the Gottwald - Melbourne test
(Gottwald and Melbourne, 2009) was threefold: it discriminates between sta-
ble, cyclic and chaotic, it escalates easily to systems of higher dimensions, its
computation is fast and it performs better than any other method we tried when

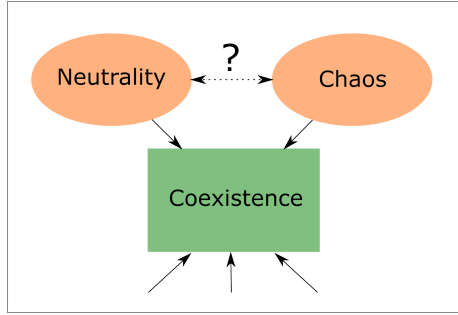


Figure 5: In our model, neutrality and chaos are not independent explanations of coexistence.

compared to the visual inspection of the time series. Additionally, we expect that the massive number of simulated ecosystems will minimize the effect of possible wrongly classified time series. For these reasons, we think that our approach to chaos detection, despite being open to improvement, suffices to detect the overall patterns.

Our results suggest a fundamentally new way in which neutrality-on-average may promote biodiversity. In addition to weakening the forces of competitive exclusion (Scheffer et al., 2018), our analyses reveal that near neutrality may boost the chances for chaotic and cyclic dynamics. As chaos and cycles may facilitate super-saturated co-existence, our findings point to a potentially widespread mechanism of maintaining biodiversity.

5 Acknowledgments

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252 **A Online appendix**

This is the Online Appendix for the paper:

254 Rodríguez-Sánchez P, van Nes EH, Scheffer M. *Neutral competition boosts
chaos in food webs.*

In the main body of the paper we focused our attention in families of food webs consisting of 12 prey and 8 predator species. In this section we show the results of the same analysis for food webs of different sizes.

A.1.1 Probability of chaos grouped by number of species

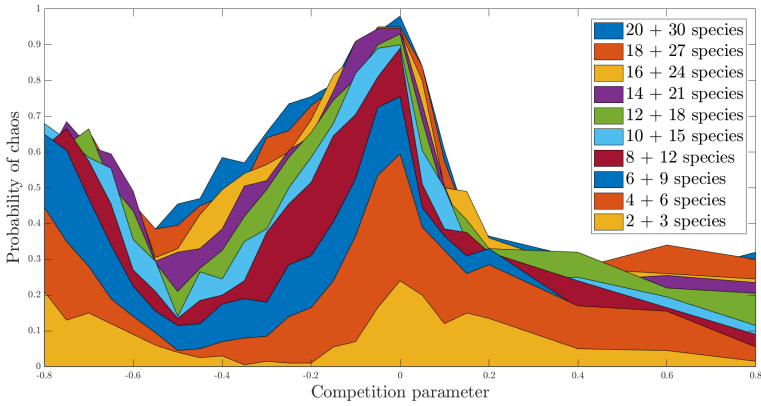


Figure A.1: Probabilities of chaos vs. competition parameter for the whole set of simulations. The competition parameter ϵ is on the horizontal axis. The estimated probability of chaos is represented on the vertical one. Each panel corresponds to an ecosystem with a different number of interacting species. The exact number is shown in each box, as number of predator + number of prey species.

A.1.2 Probability of each dynamical regime

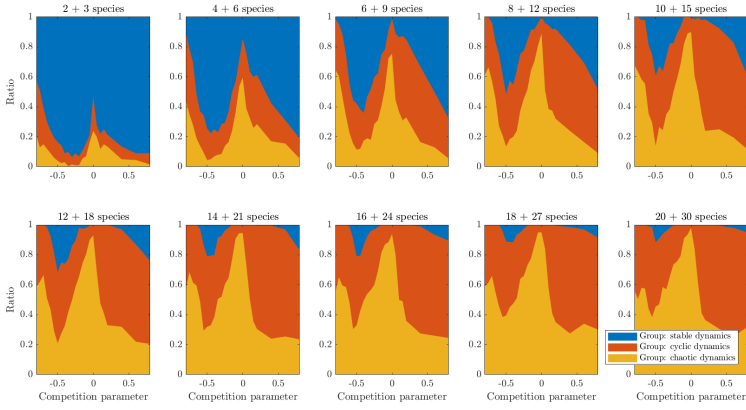


Figure A.2: Ratio of each dynamical regime vs. competition parameter for the whole set of simulations. The competition parameter ϵ is on the horizontal axis. The system size is shown in each box, as number of predator + number of prey species.

A.1.3 Biodiversity measurements

For each simulation, a biodiversity index was estimated as the number of prey species whose population was higher than a minimum threshold of 0.01 mg l^{-1} , averaged respective to time.

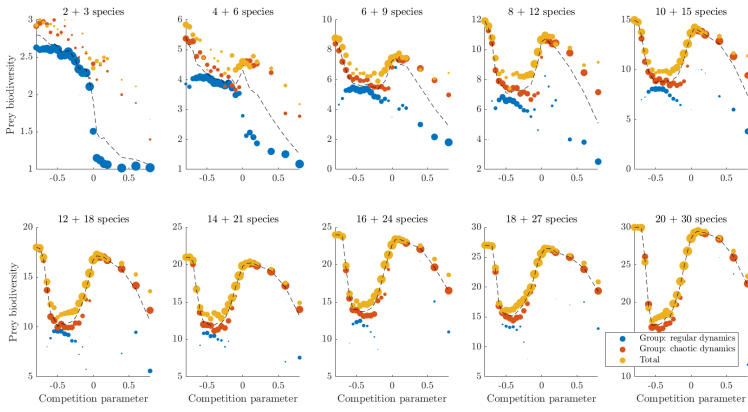


Figure A.3: Average prey biodiversity vs. competition parameter. Each panel shows a food network of a different size. For each value of the competition parameter, 200 randomly drawn ecosystems were simulated. The dashed line shows the average number of prey species of these 200 simulations. The white circles represent the average prey biodiversity of those simulations who had chaotic dynamics, the black circles represent the same for non-chaotic dynamics. The relative area of the white to the black circles represents the ratio of chaotic to regular dynamics.

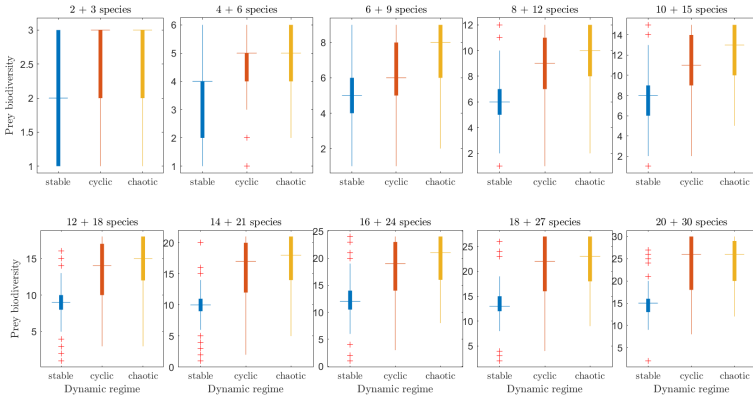


Figure A.4: Box and whisker plot of the prey biodiversity, after being classified as regular or chaotic.

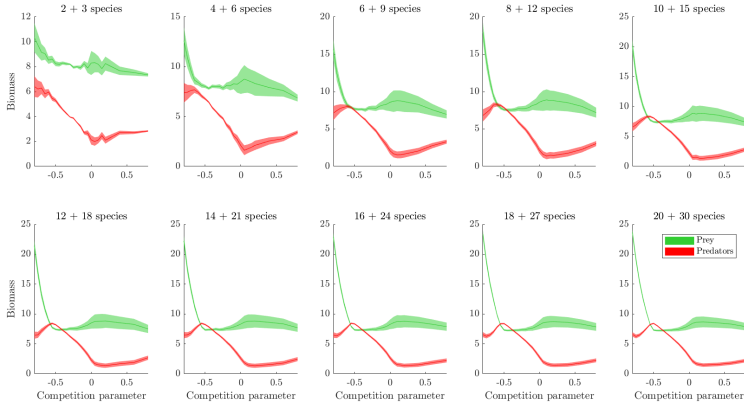


Figure A.5: Average biomasses grouped by trophic level vs. competition parameter. The width represents standard deviation.

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