

No robust coexistence in a canonical model of plant-soil feedbacks

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Abstract

Plant-soil feedbacks (PSFs) are thought to represent a crucial mechanism generating frequency-dependent dynamics in plant communities. Negative feedbacks, in particular, are routinely invoked to explain coexistence and the maintenance of diversity in species-rich communities. However, the primary modeling framework used to study PSFs considers only two plant species, and we lack clear theoretical expectations for how these complex interactions play out in communities with natural levels of diversity. Here, we demonstrate that this canonical model for PSFs is equivalent to a well-studied model from evolutionary game theory, and we use this equivalence to characterize the dynamics with an arbitrary number of plant species. Surprisingly, we find that coexistence of more than two species is virtually impossible in this model, suggesting that alternative theoretical frameworks are needed to describe feedbacks observed in diverse natural communities.

1 Introduction

It has become well understood that reciprocal interactions between plants and the soil biota, known as plant-soil feedbacks (PSFs), play an important role in structuring the composition and dynamics of plant communities. PSFs operate alongside other factors, including abiotic drivers (Bennett & Klironomos 2019) and above-ground trophic interactions (Van der Putten *et al.* 2009), but are thought to be a key mechanism generating negative frequency-dependent feedbacks that promote coexistence and maintain plant diversity (Kulmatiski *et al.* 2008; Van der Putten *et al.* 2013; Bever *et al.* 2015). The existence of PSFs has long been known (Van der Putten *et al.* 1993; Bever 1994), but our understanding of their importance – particularly in relation to patterns of coexistence – has developed rapidly in recent years (Klironomos 2002; Petermann *et al.* 2008; Mangan *et al.* 2010). Broad interest in PSFs was ignited by the development of simple mathematical models, which illustrated the potential of PSFs to mediate plant coexistence (Bever *et al.* 1997; Bever 2003; Ke & Miki 2015). These models have played a guiding role for a wide range of empirical studies, as well (Kulmatiski *et al.* 2008, 2011; Pernilla Brinkman *et al.* 2010).

The first, and still most widely known and used, model for PSFs was introduced by Bever and colleagues in the 1990s (Bever 1992; Bever *et al.* 1997; Bever 1999, 2003). In this framework, often referred to simply as the Bever model, each plant species is assumed to

31 promote the growth of a specific soil component (i.e. associated bacteria, fungi, invertebrates,
 32 considered collectively) in the vicinity of individual plants. In turn, the fitness of each plant
 33 species is impacted by the relative frequency of different soil components. Starting from
 34 minimal assumptions, Bever *et al.* (1997) derived a set of differential equations to capture
 35 these dynamics. PSFs can be either positive (fitness of a plant species is increased by its
 36 corresponding soil component) or negative (a plant species experiences lower relative fitness
 37 in its own soil). Bever *et al.* introduced a single quantity to summarize whether community-
 38 wide PSFs are positive or negative, and showed that this value characterizes the dynamical
 39 behavior of the model. In the original Bever model of two plant species, positive PSFs lead
 40 to exclusion of one species, while negative PSFs result in neutral oscillations. It is thus
 41 widely suggested that negative PSFs help sustain coexistence in real-world plant communities
 42 (Kulmatiski *et al.* 2008; Van der Putten *et al.* 2013), perhaps with spatial asynchrony playing
 43 a role in stabilizing the cyclic dynamics (Revilla *et al.* 2013; Bever 2003).

44 Subsequent studies have generalized this model to include, for example, more realistic
 45 functional forms (Umbanhowar & McCann 2005; Eppinga *et al.* 2006), more explicit repre-
 46 sentations of the soil community (Bever *et al.* 2010), spatial structure (Eppinga *et al.* 2006;
 47 Molofsky *et al.* 2002; Suding *et al.* 2013), or additional processes such as direct competitive
 48 interactions between plants (Bever 2003). However, the original Bever model remains an im-
 49 portant touchstone for the theory of PSFs (Ke & Miki 2015; Ke & Wan 2020), and informs
 50 empirical research through the interaction coefficient, I_s , derived by Bever *et al.*, which is
 51 commonly measured and used to draw conclusions about coexistence in experimental studies.
 52 Despite the ubiquity of this model, and the fruitful interplay of theory and experiment in the
 53 PSF literature, extensions to communities with more than two or three species have appeared
 54 only rarely and recently (but see Eppinga *et al.* 2018; Mack *et al.* 2019). While PSF models
 55 motivate hypotheses and conclusions about species-rich natural communities, there is much
 56 still unknown about the behavior of these models with natural levels of diversity (Van der
 57 Putten *et al.* 2013).

58 Here, we extend the Bever model to include any number of plant species, and show that
 59 the model is equivalent to a special form of the replicator equation studied in evolutionary
 60 game theory (Hofbauer & Sigmund 1998). In particular, this model corresponds to the class of
 61 bimatrix games, where there are two players (here, plants and soil components) which interact

with asymmetric strategies and payoffs. The replicator dynamics of bimatrix games are well-studied, allowing us to characterize many properties of the Bever model with n plant species. Surprisingly, using this equivalence, we show that coexistence of more than two species in this model is never robust.

2 Results

2.1 Generalizing a classic PSF model

Inspired by emerging empirical evidence for the important role of PSFs in plant community dynamics and coexistence (Van der Putten *et al.* 1993; Bever 1994), Bever *et al.* (1997) introduced a simple mathematical model to investigate their behavior. In this model, two plant species, 1 and 2, grow exponentially with growth rates determined by the state of the soil biota in the system. These effects of soil on plants are specified by parameters α_{ij} , the growth rate of plant species i in soil type j . There is a soil component corresponding to each plant species, which grows exponentially in the presence of its associated plant at a rate β_i . Bever *et al.* set an important precedent by considering dynamics of *relative* abundances in such a system; starting from dynamics of the form

$$\begin{cases} \frac{dx_i}{dt} = x_i \left(\frac{\alpha_{ii} y_i + \alpha_{ij} y_j}{y_i + y_j} \right), & i, j = 1, 2 \\ \frac{dy_i}{dt} = y_i \left(\frac{\beta_i x_i}{x_i + x_j} \right) \end{cases} \quad (1)$$

for the *absolute* abundances of plants (x_i) and soil components (y_i), one considers the relative abundances (frequencies), $p_i = x_i / \sum_j x_j$ and $q_i = y_i / \sum_j y_j$. The dynamics for these frequencies are easily derived from Eq. 1, and using the facts $p_i = 1 - p_j$ and $q_i = 1 - q_j$, can be written as:

$$\begin{cases} \frac{dp_i}{dt} = p_i p_j ((\alpha_{ii} - \alpha_{ji}) q_i + (\alpha_{ij} - \alpha_{jj}) q_j), & i, j = 1, 2 \\ \frac{dq_i}{dt} = q_i q_j (\beta_i p_i - \beta_j p_j). \end{cases} \quad (2)$$

This model may admit a coexistence equilibrium where

$$\begin{cases} p_i^* &= \frac{\beta_j}{\beta_i + \beta_j}, \quad i, j = 1, 2 \\ q_i^* &= \frac{\alpha_{jj} - \alpha_{ji}}{\alpha_{ii} - \alpha_{ij} + \alpha_{jj} - \alpha_{ji}}. \end{cases} \quad (3)$$

82 A central finding of the analysis by Bever *et al.* was that the denominator of q_i^* , which they
 83 termed the “interaction coefficient”, I_s ($= \alpha_{ii} - \alpha_{ij} + \alpha_{jj} - \alpha_{ji}$), controls the model dynamics:
 84 When $I_s > 0$, which represents a community with positive feedbacks, the equilibrium in
 85 Eq. 3 is unstable, and the two species cannot coexist. On the other hand, when $I_s < 0$,
 86 the equilibrium is neutrally stable, and the dynamics cycle around it, providing a form of
 87 non-equilibrium coexistence. In fact, these conclusions depend on the existence of a feasible
 88 equilibrium (i.e. positive equilibrium values), which further requires that $\alpha_{ii} < \alpha_{ji}$ for both
 89 $i, j = 1, 2$, in order for the model to exhibit coexistence (Bever *et al.* 1997; Ke & Miki 2015).

90 This coexistence is fragile. Species frequencies oscillate neutrally, similar to the textbook
 91 example of Lotka-Volterra predator-prey dynamics. Any stochasticity, external forcing, or
 92 time variation in the model parameters can destroy these finely balanced oscillations and
 93 cause one species to go extinct (Revilla *et al.* 2013). However, coupled with mechanisms
 94 that buffer the system from extinctions, such as migration between desynchronized patches
 95 or the presence of a seed bank, the negative feedbacks in this model might produce sustained
 96 coexistence (Revilla *et al.* 2013; Bever 2003).

97 Of course, most natural plant communities feature more than two coexisting species, and it
 98 is precisely in the most diverse communities that mechanisms of coexistence hold the greatest
 99 interest (Van der Putten *et al.* 2013). While it is not immediately clear how to generalize
 100 Eq. 2 to more than two species, Eq. 1 is naturally extended by maintaining the assumption
 101 that the overall growth rate for any plant is a weighted average of its growth rate in each soil
 102 type:

$$\begin{cases} \frac{dx_i}{dt} &= x_i \left(\sum_j \alpha_{ij} q_j \right), \quad i = 1, \dots, n \\ \frac{dy_i}{dt} &= y_i (\beta_i p_i). \end{cases} \quad (4)$$

103 From Eq. 4, one can derive the n -species analogue of Eq. 2,

$$\begin{cases} \frac{dp_i}{dt} = p_i \left(\sum_j \alpha_{ij} q_j - \sum_{j,k} \alpha_{jk} p_j q_k \right), & i = 1, \dots, n \\ \frac{dq_i}{dt} = q_i \left(\beta_i p_i - \sum_j \beta_j p_j q_j \right) \end{cases} \quad (5)$$

104 giving the dynamics for species and soil component frequencies (Fig. 1). Eq. 5 is conveniently
 105 expressed in matrix form as

$$\begin{cases} \frac{d\mathbf{p}}{dt} = D(\mathbf{p}) (A\mathbf{q} - (\mathbf{p}^T A \mathbf{q}) \mathbf{1}) \\ \frac{d\mathbf{q}}{dt} = D(\mathbf{q}) (B\mathbf{p} - (\mathbf{q}^T B \mathbf{p}) \mathbf{1}) \end{cases} \quad (6)$$

106 where vectors are indicated in boldface (e.g. \mathbf{p} is the vector of species frequencies $(p_1, p_2, \dots, p_n)^T$
 107 and $\mathbf{1}$ is a vector of n ones) and $D(\mathbf{z})$ is the diagonal matrix with vector \mathbf{z} on the diagonal.
 108 We have introduced the matrices $A = (\alpha_{ij})$ and $B = D(\beta_1, \beta_2, \dots, \beta_n)$, specifying soil effects
 109 on plants and plant effects on soil, respectively. Because \mathbf{p} and \mathbf{q} are vectors of frequencies,
 110 they must sum to one: $\mathbf{1}^T \mathbf{p} = \mathbf{1}^T \mathbf{q} = 1$. Using these constraints, one can easily show that the
 111 Bever model (Eq. 2) is a special case of Eqs. 5 and 6 when $n = 2$.

112 2.2 Equivalence to bimatrix game dynamics

113 Systems that take the form of Eq. 6 are well-known and well-studied in evolutionary game
 114 theory. Our generalization of the Bever model is a special case of the *replicator equation*,
 115 corresponding to the class of *bimatrix games* (Taylor 1979; Hofbauer 1996; Hofbauer & Sig-
 116 mund 1998; Cressman & Tao 2014). Bimatrix games arise in diverse contexts, such as animal
 117 behavior (Taylor 1979; Selten 1988), evolutionary theory (Hofbauer & Sigmund 1998; Cress-
 118 man & Tao 2014), and economics (Friedman 1991), where they model games with asymmetric
 119 players. In a bimatrix game, each player (here, the plant community and the soil) has a dis-
 120 tinct set of strategies (plants species and soil components, respectively) and payoffs (realized
 121 growth rates).

122 Much is known about bimatrix game dynamics, and we can draw on this body of knowledge
 123 to characterize the behavior of the Bever model with n species. Essential mathematical

background and details are presented in the Supplemental Methods; for a detailed introduction to bimatrix games, see Hofbauer & Sigmund (1998).

Under the mild condition that matrix A is invertible, Eq. 6 admits a unique coexistence equilibrium given by $\mathbf{p}^* = k_p B^{-1} \mathbf{1}$ and $\mathbf{q}^* = k_q A^{-1} \mathbf{1}$, where $k_p = 1/(\mathbf{1}^T B^{-1} \mathbf{1})$ and $k_q = 1/(\mathbf{1}^T A^{-1} \mathbf{1})$ are constants of proportionality that ensure the equilibrium frequencies sum to one for both plants and soil. Because B is a diagonal matrix, and all β_i are assumed positive, the equilibrium plant frequencies, \mathbf{p}^* , are always positive, as well. Thus, feasibility of the equilibrium hinges on the soil frequencies, \mathbf{q}^* , which are all positive if the elements of $A^{-1} \mathbf{1}$ all share the same sign.

As we have seen, when the community consists of two species, the coexistence equilibrium, if feasible, can be either unstable or neutrally stable. In fact, the same is true for the n -species extension (and, more generally, for any bimatrix game dynamics, Eshel *et al.* 1983; Selten 1988; Hofbauer & Sigmund 1998). This can be established using straightforward local stability analysis, after accounting for the relative abundance constraints, which imply $p_n = 1 - \sum_{i=1}^{n-1} p_i$ and $q_n = 1 - \sum_{i=1}^{n-1} q_i$. Using these substitutions, Eq. 5 can be written as a system of $2n - 2$ (rather than $2n$) equations, and the community matrix for this reduced model has a very simple form (see Supplemental Methods). In particular, due to the bipartite structure of the model, the community matrix has all zero diagonal elements, which implies that the eigenvalues of this matrix sum to zero. These eigenvalues govern the stability of the coexistence equilibrium, and this property leaves two qualitatively distinct possibilities: either the eigenvalues have a mix of positive and negative real parts (in which case the equilibrium is unstable), or the eigenvalues all have zero real part (in which case the equilibrium is neutrally stable). Already, we can see that the model never exhibits equilibrium coexistence, regardless of the number of species.

Another notable property of bimatrix game dynamics is that the vector field defined by the model equations is divergence-free or incompressible (see Hofbauer & Sigmund 1998, for a proof). The divergence theorem from vector calculus (Arfken 1985) then dictates that Eq. 6 cannot have any attractors – that is, regions of the phase space that “pull in” trajectories – with multiple species. This rules out coexistence in a stable limit cycle or other non-equilibrium attractors (e.g. chaotic attractors). Thus, only the relatively fragile coexistence afforded by neutral oscillations is possible, as in the two-species model.

Based on the local stability properties of the coexistence equilibrium, Bever *et al.* concluded that such neutral cycles arise for two species when $\alpha_{11} < \alpha_{21}$ and $\alpha_{22} < \alpha_{12}$. The equivalence between their model and a bimatrix game with two strategies allows us to give a fuller picture of these cycles. Namely, we can identify a constant of motion for the two-species dynamics:

$$H = (\alpha_{12} - \alpha_{22}) \log q_1 + (\alpha_{21} - \alpha_{11}) \log q_2 + \beta_2 \log p_1 + \beta_1 \log p_2. \quad (7)$$

Using the chain rule and time derivatives in Eq. 2, it is easy to show that $\frac{dH}{dt} = 0$ for any plant and soil frequencies (see Supplemental Methods). The level curves of H form closed orbits around the equilibrium when the equilibrium is neutrally stable. Thus, H implicitly defines the trajectories of the model, and can be used to determine characteristics such as the amplitude of oscillations arising from particular initial frequencies (Volterra 1926).

Because neutral cycles provide the only possible form of coexistence in this model, a key question becomes whether and when neutral cycles with n species can arise. Do the “negative feedback” conditions identified by Bever *et al.* generalize in richer communities? Indeed, they do; however, for more than two species, these conditions are very severe. The model in Eq. 6 supports oscillations with n species – for any n – if matrices A and B satisfy a precise relationship (see Supplemental Methods for details). In particular, the model parameters must satisfy the conditions $\alpha_{ij} = \gamma_i + \delta_j$ for some constants γ_i, δ_i in $i = 1, \dots, n$ (when $i \neq j$), and $\alpha_{ii} = \gamma_i + \delta_i - c\beta_i$ (where c is a positive constant independent of i). In the language of bimatrix games, such systems are called *rescaled zero-sum games* (Hofbauer 1996; Hofbauer & Sigmund 1998). It is a long-standing conjecture in evolutionary game theory that these parameterizations are the *only* cases where n -species coexistence can occur (Hofbauer 1996, 2011).

Ecologically, these conditions mean there is a fixed effect of each soil type and plant species identity, and the growth rate of plant i in soil type j is the additive combination of these two, with no interaction effects. The only exception is for plants growing in their own soil type, which must experience a fitness cost ($\gamma_i + \delta_i - \alpha_{ii}$) exactly proportional to the rate at which they promote growth of their soil type (β_i). These conditions clearly extend the intuitive notion that each plant must have a disadvantage in its corresponding soil type to

allow coexistence. But the parameters of the model are constrained so strongly that we never expect to observe cycles with more than two species in practice. When $n > 2$, a great deal of fine-tuning is necessary to satisfy the conditions yielding a rescaled zero-sum game; the probability that random parameters will be suitable is infinitesimally small. We confirm this numerically with simulations shown in Fig. 2. Although n -species cycles are clearly possible (as in Fig. 3), for parameters drawn independently at random, communities always collapse to one or two species, regardless of the initial richness.

Not only are parameter combinations permitting many-species oscillations rare, they are also extremely sensitive to small changes to the parameter values. The rescaled zero-sum condition imposes many exact equality constraints on the matrix A (e.g. $\alpha_{ij} - \alpha_{ik} = \alpha_{lj} - \alpha_{lk}$ for all i, j, k , and l). Even if mechanisms exist to generate the requisite qualitative patterns, inevitable quantitative variation in real-world communities will disrupt coexistence (Fig. 3). Coexistence of $n > 2$ species – even in the weak sense of neutral cycles – is not robust to small changes in the model parameters.

Interestingly, the two-species model is not subject to the same fragility. It can be shown (see Supplemental Methods) that all 2×2 bimatrix games take the same general form as a rescaled zero-sum game, although the constant c may be positive or negative, depending on the parameters. When I_s , the interaction coefficient identified by Bever *et al.*, is negative, c is positive, ensuring (neutral) stability. This condition amounts to an inequality constraint, rather than an equality constraint, and so it *is* generally robust to small variations in model parameters (Fig. 3). As we can now see, the case $n = 2$ is unique in this regard.

3 Discussion

The Bever model has played a central role in motivating PSF research, and continues to guide both theory and experiment in this fast-growing field (Bever *et al.* 2015; Kandlikar *et al.* 2019; Ke & Wan 2020). Here, we extend the Bever model to any number of species, and highlight its equivalence to bimatrix game dynamics. Taking advantage of the well-developed theory for these dynamics, we are able to characterize the behavior of the generalized Bever model in detail.

Our central finding is that there can be no robust coexistence of plant species in this model.

212 Regardless of the number of species, n , the model never exhibits equilibrium coexistence or
213 other attractors. Coexistence can be attained through neutral oscillations, but these dynamics
214 lack any restoring force and are easily destabilized by stochasticity or exogenous forcing. In
215 this respect, the generalized model behaves similarly to the now-classic two-species system.
216 However, unlike the two-species model, oscillations with $n > 2$ species can only result under
217 very restricted parameter combinations. These parameterizations are vanishingly unlikely to
218 arise by chance, and highly sensitive to small deviations. Thus, coexistence of more than two
219 species is neither dynamically nor structurally stable.

220 This result may be surprising, because a significant body of experimental evidence indi-
221 cates that PSFs can play an important role in mediating the coexistence of more than two
222 species in natural communities (Kulmatiski *et al.* 2008; Petermann *et al.* 2008; Mangan *et al.*
223 2010; Bever *et al.* 2015). Apparently, the picture suggested by the two-species Bever model
224 generalizes in nature, but not in the model framework itself. We note that this framework was
225 introduced as an intentional simplification to illustrate the potential role of PSFs in mediating
226 coexistence, not to accurately model the biological details of PSFs. Indeed, the model has
227 been wildly successful in spurring research into PSFs. Alongside empirical study of these
228 processes, other modeling approaches have emerged, accounting for more biological realism
229 (e.g., Umbanhowar & McCann 2005; Eppstein & Molofsky 2007; Bever *et al.* 2010), or with
230 the demonstrated capacity to produce multispecies coexistence (e.g., Bonanomi *et al.* 2005;
231 Miller & Allesina 2021). Some of these are minor modifications of the Bever model framework;
232 others build on distinct foundations (Ke & Miki 2015; Ke & Wan 2020). Our results suggest
233 that these various avenues are worth pursuing further.

234 Alternative modeling approaches are particularly important for better integration of the-
235 ory and data. The predictions of the Bever model are commonly used to guide the design
236 and analysis of PSF experiments, especially in drawing conclusions about coexistence. Our
237 analysis cautions that applications of this model in multispecies communities might lead to
238 incorrect inference. For example, attempts to parameterize the Bever model for three species
239 using empirical data have produced predictions of non-coexistence in plant communities that
240 coexist experimentally (Kulmatiski *et al.* 2011). In many other studies, the pairwise in-
241 teraction coefficient, I_s , is calculated for species pairs and used to assess whole-community
242 coexistence (Kulmatiski *et al.* 2008; Fitzsimons & Miller 2010; Pendergast IV *et al.* 2013; Sud-

ing *et al.* 2013; Kuebbing *et al.* 2015; Smith & Reynolds 2015; Bauer *et al.* 2017; Kandlikar
et al. 2019). However, we have seen that whole-community coexistence is virtually impossible
within the generalized model, and there is no guarantee that the pairwise coexistence condi-
tions for this model will agree with n -species coexistence conditions in other frameworks. For
example, $I_s < 0$ for all species pairs is neither necessary nor sufficient to produce coexistence
in a metapopulation-based model for PSFs (Miller & Allesina 2021).

In contrast to other extensions of the Bever model to many-species communities (Eppinga
et al. 2018; Mack *et al.* 2019), our approach keeps the dynamics of both plants and soil fully
explicit. As such, we make no assumptions about the relative timescales of plant and soil
dynamics, or whether either of these reach equilibrium. This difference likely explains the
discrepancy between our conclusions and previously published predictions for n -species PSFs
(Kulmatiski *et al.* 2011; Eppinga *et al.* 2018; Mack *et al.* 2019).

PSFs as envisioned in the classic Bever model might facilitate coexistence in conjunction
with other mechanisms, such as direct species interactions (Bever 2003), or through long tran-
sient dynamics, but our analysis shows that they cannot produce robust n -species coexistence
in isolation. This finding calls for renewed theoretical investigation of PSFs. One important
consideration is grounding PSF modeling frameworks in more realistic models for absolute
abundances or densities. As various researchers have noted (Kulmatiski *et al.* 2011; Revilla
et al. 2013; Eppinga *et al.* 2018; Ke & Wan 2020), the Bever model and its extensions are in
fact projections (onto the space of relative abundances, or frequencies) of dynamics for plant
and soil *abundances*. Consequently, the projected dynamics can mask unbiological outcomes
in the original model (e.g. relative abundances oscillate around equilibrium while absolute
abundances shrink to zero or explode to infinity). Indeed, the absolute abundance model
(Eq. 4) used to derive our n -species frequency dynamics (Eqs. 5-6) does not generally possess
any fixed points, which is a basic requirement for species coexistence (Hutson 1990; Hutson &
Schmitt 1992). It is usually seen as desirable to study PSFs in the space of species frequencies,
both because this facilitates connections to data, and because frequencies are considered a
more appropriate metric for analyzing processes that stabilize coexistence (Adler *et al.* (2007);
Eppinga *et al.* (2018), but see Kandlikar *et al.* (2019); Ke & Wan (2020)). But models that
introduce frequencies through a natural constraint, such as competition for finite space, will
likely produce more realistic dynamics.

274 From a broader theoretical perspective, the qualitative change in model behavior that we
 275 observe as the number of species increases from two to three or more is a striking phenomenon,
 276 but not an unprecedented one. Ecologists have repeatedly found that intuitions from two-
 277 species models can generalize (or fail to generalize) to more diverse communities in surprising
 278 ways (Strobeck 1973; Smale 1976; Barabás *et al.* 2016). Our analysis provides another illus-
 279 tration of the fact that “more is different” (Anderson 1972) in ecology, and highlights the
 280 importance of developing theory for species-rich communities.

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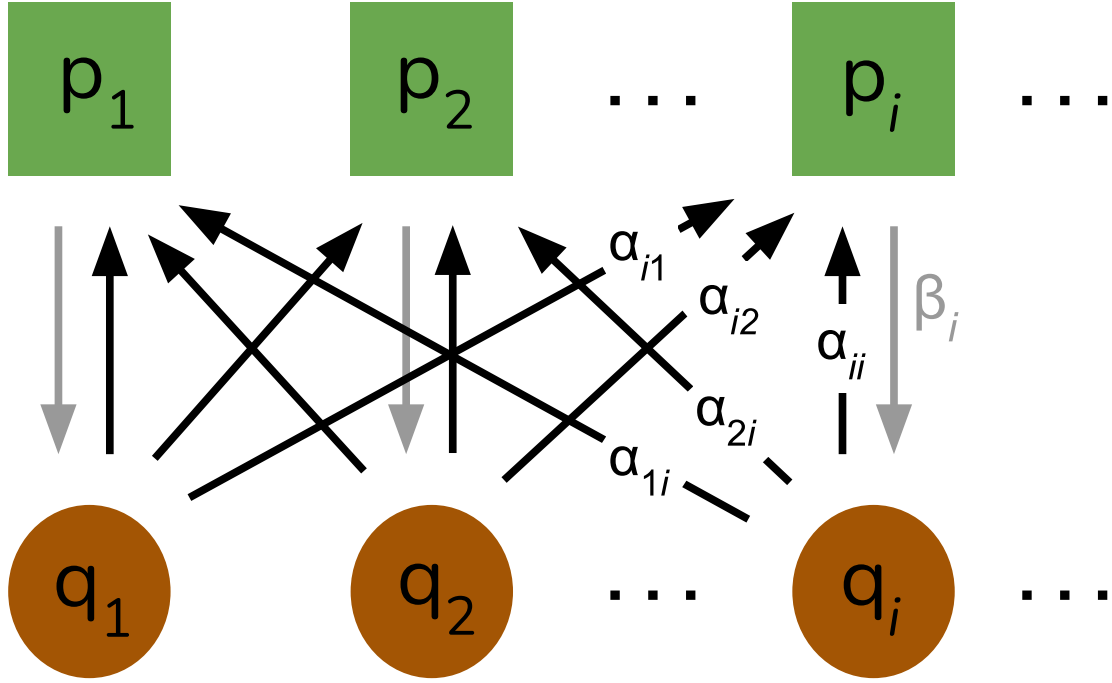


Figure 1: The model described by Eqs. 5-6 is shown here graphically. Plant species (green squares) promote the growth of their respective soil components (brown circles) at a rate β_i (gray arrows). In turn, growth of each plant is governed by the mix of soil components present in the system, with the effect of soil component j on species i quantified by the parameter α_{ij} (black arrows). This model is a straightforward extension of the model proposed by Bever *et al.* (1997) to an arbitrary number of species. Note only selected parameter labels are shown for clarity.

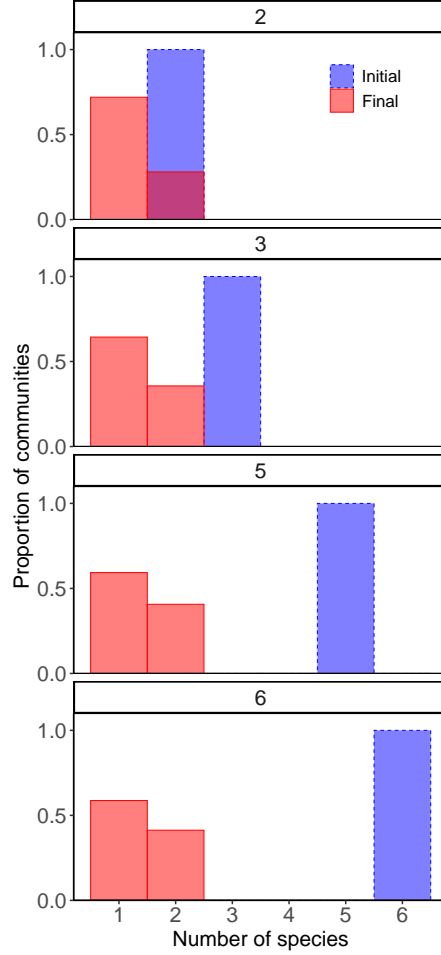


Figure 2: Final community sizes with varying initial richness. We show the distribution of final richness (number of species, in red) for 5000 communities governed by the n -species Bever model, initialized with 2, 3, 5, or 6 plant species. Parameters α_{ij} and β_i were sampled independently from a standard uniform distribution, $U(0, 1)$. For each random parameterization at each level of initial richness, we integrated the dynamics of Eq. 6 until the system reached a periodic orbit or until only one species remained. In agreement with the conjecture that coexistence of more than two species is vanishingly unlikely, we found that regardless of the number of species initially present, every community collapsed to a subset with one or two surviving species.

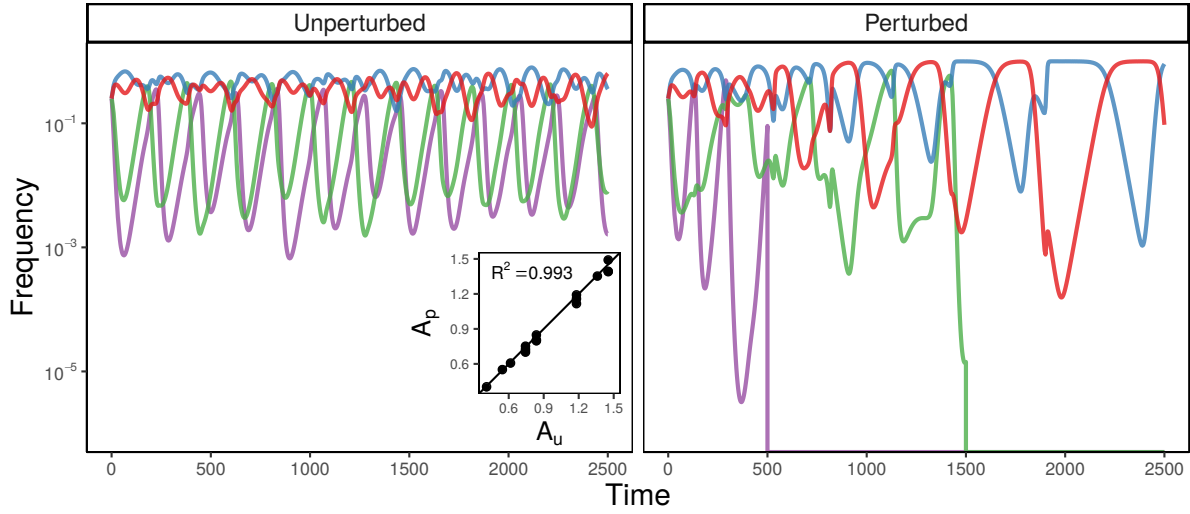


Figure 3: Coexistence of three or more species is not robust. It is possible to obtain neutrally stable oscillations with any number of species if the model parameters constitute a rescaled zero-sum game (see text for conditions). Here, for example, we show sustained oscillations with 4 species using fine-tuned parameters, A_u (left). However, if we randomly perturb A_u by a small amount to obtain new parameters, A_p , the dynamics quickly collapse to a two-species subset (right). Any slight perturbation is enough to disrupt coexistence; for this example, the parameters A_u and A_p are highly correlated (inset) and differ in value by less than 3% on average.