

ARTICLE

Does deterministic coexistence theory matter in a finite world?

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Abstract

Contemporary studies of species coexistence are underpinned by deterministic models that assume that competing species have continuous (i.e., noninteger) densities, live in infinitely large landscapes, and coexist over infinite time horizons. By contrast, in nature, species are composed of discrete individuals subject to demographic stochasticity and occur in habitats of finite size where extinctions occur in finite time. One consequence of these discrepancies is that metrics of species' coexistence derived from deterministic theory may be unreliable predictors of the duration of species coexistence in nature. These coexistence metrics include invasion growth rates and niche and fitness differences, which are now commonly applied in theoretical and empirical studies of species coexistence. In this study, we tested the efficacy of deterministic coexistence metrics on the duration of species coexistence in a finite world. We introduce new theoretical and computational methods to estimate coexistence times in stochastic counterparts of classic deterministic models of competition. Importantly, we parameterized this model using experimental field data for 90 pairwise combinations of 18 species of annual plants, allowing us to derive biologically informed estimates of coexistence times for a natural system. Strikingly, we found that for species expected to deterministically coexist, community sizes containing only 10 individuals had predicted coexistence times of more than 1000 years. We also found that invasion growth rates explained 60% of the variation in intrinsic coexistence times, reinforcing their general usefulness in studies of coexistence. However, only by integrating information on both invasion growth rates and species' equilibrium population sizes could most (>99%) of the variation in species coexistence times be explained. This integration was achieved with demographically uncoupled single-species models solely determined by the invasion growth rates and equilibrium population sizes. Moreover, because of a complex relationship between niche overlap/fitness differences and equilibrium population sizes, increasing niche overlap and increasing fitness differences did not always result in decreasing coexistence times, as deterministic theory would predict. Nevertheless, our results tend to support the informed use of deterministic theory for understanding the duration of species' coexistence while

highlighting the need to incorporate information on species' equilibrium population sizes in addition to invasion growth rates.

KEYWORDS

annual plants, coexistence, competition, demographic stochasticity, extinction, modern coexistence theory

INTRODUCTION

Understanding how competing species coexist is a central problem in ecology (Chesson, 2000a; Hutchinson, 1961). Recent theoretical progress on this problem has replaced vague conceptualizations of coexistence requirements with tools that allow ecologists to quantify the niche differences that allow coexistence and the fitness differences that drive competitive exclusion (Adler et al., 2007; Chesson, 1990, 2000a). This progress has, in turn, motivated a large number of empirical studies to apply tools derived from theory to quantify the drivers of species coexistence in the field (e.g., Levine & Lambers, 2009; Narwani et al., 2013). Importantly, however, the deterministic theory on which these advances are based assumes that competition occurs between populations whose densities vary continuously on landscapes of infinite size (Faure & Schreiber, 2014; Schreiber, 2017). Under these assumptions, the influence of processes occurring as a consequence of the discrete nature of individuals—such as demographic stochasticity—are excluded (Hart et al., 2016; Pande, Fung, Chisholm, & Shnerb, 2020). This generates a fundamental disconnect between theory and reality because, although theory predicts that coexisting species will coexist indefinitely, in nature coexistence can only occur over finite periods of time. How well metrics derived from deterministic coexistence theory predict the duration of coexistence in the discrete, finite systems of nature—the ultimate object of study—remains largely unknown.

One of the most widely used metrics in contemporary theoretical and empirical studies of species coexistence is invasion growth rate (Chesson, 2000a; Grainger et al., 2019; Hofbauer & Sigmund, 1998; Schreiber, 2000). A metric from deterministic models, the invasion growth rate of a species is its per-capita growth rate at vanishingly low densities when its competitors' densities are at equilibrium. For two competing species, deterministic models predict that coexistence occurs when each species in a competing pair has a positive long-term invasion growth rate (MacArthur & Levins, 1967). Provided there are no Allee effects or positive frequency dependence at low densities (Schreiber et al., 2019), meeting this “mutual-invasibility” criterion implies that coexistence occurs indefinitely. Invasion analyses have been

particularly powerful in studies of species coexistence because model-specific expressions for the invasion growth rate can be used to derive expressions that quantify the magnitude of niche and fitness differences between species (Chesson, 1990, 2013; Hart et al., 2018). Thus, the mutual invasibility criterion has become central to much of our current understanding of species coexistence. However, when finite populations with positive deterministic invasion growth rates are depressed to low numbers of discrete individuals, they may still fail to persist because the negative effects of demographic stochasticity lead to extinction (Hart et al., 2018; Pande, Fung, Chisholm, & Shnerb, 2020). More generally, even when species have higher densities away from the invasion boundary, demographic stochasticity operating on competing populations of finite size ensures extinction in finite time (Jagers, 2010; Lande et al., 2003; Reuter, 1961; Schreiber, 2017).

Despite the dominance of coexistence theory that imposes continuously varying population densities and infinitely large landscapes on a discrete, finite world, the effect of demographic stochasticity on coexistence is receiving greater attention (Shoemaker et al., 2020; Vellend, 2010). Perhaps the most prominent example is neutral theory, in which demographic stochasticity is the sole driver of competitive dynamics between species (Hubbell, 2001; Hubbell & Foster, 1986). Like fixation times for neutral alleles (Ewens, 2012), neutral theory predicts that competing species can coexist for long periods of time when the community size is large relative to the number of species (Hubbell, 2001). The well-known problem with neutral theory, however, is that in emphasizing the role of demographic stochasticity, the theory simultaneously excludes all deterministic processes governing species interactions (Vellend, 2010). This exclusion includes species-level fitness differences that drive competitive exclusion and niche differences that promote coexistence (Adler et al., 2007). Thus, depending on the relative strength of these deterministic processes, coexistence times predicted by neutral theory (i.e., by the action of demographic stochasticity alone) will be either significant over- or underestimates.

More recently, population theoretic studies that incorporate demographic stochasticity into traditionally

deterministic models of competition have begun to emerge (Adler & Drake, 2008; Gabel et al., 2013; Gómez-Corral & López García, 2012; Kramer & Drake, 2014). These studies first make the important point that interspecific competitive interactions cause extinction dynamics to be different from that predicted by single-species models, and they also demonstrate that stochasticity causes the identity of the winner in competition to be less than perfectly predicted by deterministic competition-model parameters (Gabel et al., 2013; Gómez-Corral & López García, 2012; Kramer & Drake, 2014). In addition, and together with models of extinction processes applied to single-species dynamics, these studies also highlight the likely importance of variables not traditionally considered in contemporary studies of species coexistence (Gómez-Corral & López García, 2012; Kramer & Drake, 2014). For example, in single-species models, time to extinction depends critically on equilibrium population size (Boyce, 1992; Grimm & Wissel, 2004; Ovaskainen & Meerson, 2010). If equilibrium population size is similarly important for the duration of species' coexistence, then the sole focus on invasion growth rates as the primary arbiter of coexistence may be problematic.

Despite recent progress in this area, existing studies of the effects of demographic stochasticity on the duration of coexistence concentrate only on cases where competitive exclusion is deterministically ensured (Gómez-Corral & López García, 2012; Kramer & Drake, 2014). What is missing, therefore, are assessments of the duration of coexistence when coexistence rather than exclusion is deterministically ensured and how these durations relate to existing deterministic coexistence metrics (e.g., invasion growth rates, niche and fitness differences). The latter point is particularly important as long as the field continues to rely on deterministic theory to identify coexistence mechanisms and to interpret empirical results (Ellner et al., 2019). Indeed, Pande, Fung, Chisholm, and Shnerb (2020) identify inadequacies in invasion growth rates derived from deterministic theory for predicting coexistence times of finite populations in fluctuating environments. They show that the environmentally dependent distribution of species' growth rates in fluctuating environments—not just the average “invasion growth rate”—influences coexistence times. Although this finding is important, it leaves the independent effects of environmental and demographic stochasticity on coexistence times unresolved. For example, for coexistence mechanisms not reliant on environmental fluctuations there is no distribution of environmentally dependent population growth rates, and so it remains unclear how demographic stochasticity alone influences coexistence times. Moreover, there is considerable empirical attention on, and support for, the

contribution to coexistence of fluctuation-independent coexistence mechanisms (Adler et al., 2010; Chu & Adler, 2015; Ellner et al., 2016; Godoy & Levine, 2014; Letten et al., 2017; Li et al., 2019; Mordecai et al., 2016; Muller-Landau & Visser, 2019; Spaak & De Laender, 2020; Wainwright et al., 2019; Zepeda & Martorell, 2019). For example, Zepeda and Martorell (2019) found that the coexistence of 17 grassland species was mostly due to large fluctuation-independent coexistence mechanisms. Hence, it is particularly important to understand the applicability of these fluctuation-independent mechanisms for communities influenced by demographic stochasticity in nature.

In this paper, we explore the duration of species coexistence in discrete, finite natural systems experiencing demographic stochasticity. In particular, we determine the relationship between the duration of coexistence in the presence of demographic stochasticity and commonly used metrics of coexistence derived from deterministic theory. To determine these relationships, we introduce the concept of the intrinsic coexistence time, a multispecies analogue of Grimm and Wissel's (2004) intrinsic extinction time for single-species stochastic population models. This metric corresponds to the mean time to losing one or more of these species after the species were coexisting sufficiently long to exhibit relatively stationary population dynamics, that is, quasi-stationarity. Our assessment of this metric is based on novel mathematical and computational methods that allow us to derive explicit relationships between the quasi-stationary behavior of a stochastic model and the dynamics of its deterministic model counterpart (Faure & Schreiber, 2014). Importantly, we ground our analytical approach using estimates of competition model parameters from 90 pairs of annual plant species competing on serpentine soils in the field (Godoy et al., 2014).

MODELS AND METHODS

Models and methods overview

Our models and methods are composed of two broad parts. First, we describe the model and its empirical parameterization, introduce metrics of deterministic coexistence, and describe new analytical methods for calculating coexistence times in the presence of demographic stochasticity. Second, we describe our methods for answering a series of questions about the relationship between deterministic coexistence and stochastic coexistence times using the empirically parameterized model.

Model

We base our analysis on an annual plant competition model (Beverton & Holt, 1957; Leslie & Gower, 1958; Watkinson, 1980), which is well studied analytically (Cushing et al., 2004) and does a good job of describing competitive population dynamics in plant communities in the field (Godoy & Levine, 2014). In the deterministic model with no demographic stochasticity, the dynamics of species i ($i = 1$ or 2) can be expressed in terms of its density $n_{i,t}$ and its competitor's density $n_{j,t}$ at time t :

$$n_{i,t+1} = \frac{\lambda_i n_{i,t}}{1 + (\alpha_{ii} n_{i,t} + \alpha_{ij} n_{j,t})(\lambda_i - 1)}, \text{ where } j \neq i, \quad (1)$$

where λ_i describes offspring production in the absence of competition, and α_{ii} and α_{ij} are the competition coefficients, which describe the rate of decline in offspring production as conspecific and heterospecific competitor densities increase, respectively. Including the multiplicative factor $\lambda_i - 1$ in the denominator ensures that the competition coefficients are in units of proportional reductions in λ_i , such that when $\alpha_{ii} n_i + \alpha_{ij} n_j = 1$, the population stops growing. Parameterizing our model in this way ensures that the conditions for coexistence in our model will correspond to the classical conditions for coexistence in the continuous-time Lotka–Volterra competition model. In the deterministic model given by Equation (1), offspring production and population densities take on values in the nonnegative, real numbers.

The stochastic model takes the same functional form, but, in contrast to the deterministic model, offspring production and population sizes $N_{i,t}$ take on nonnegative integer values. Population densities $N_{i,t}/S$ are determined by the community size S and take on nonnegative rational values. Specifically, in the stochastic model, discrete individuals produce random numbers of discrete offspring according to a Poisson distribution. These random individual-level reproductive events generate demographic stochasticity in our model. Assuming the Poisson-distributed offspring production of individuals are independent with the same mean as the deterministic model, the sum of these events is also Poisson distributed, and the dynamics of our stochastic model at the population level can be expressed as

$$N_{i,t+1} = \text{Pois} \left(\frac{\lambda_i N_{i,t}}{1 + (\alpha_{ii} N_{i,t}/S + \alpha_{ij} N_{j,t}/S)(\lambda_i - 1)} \right) \quad (2)$$

where $\text{Pois}(\mu)$ denotes a Poisson random variable with mean μ .

The community size parameter S allows us to quantify the effects of demographic stochasticity on coexistence in

landscapes of finite size. Importantly, when community size S is sufficiently large, the dynamics of the densities $N_{i,t}/S$ of our stochastic model given by Equation (2) are well approximated by the deterministic model given by Equation (1) (Figure 1a–c). This justifies our analysis of the effects of metrics of deterministic coexistence on stochastic coexistence times. Specifically, for any prescribed time interval, say $[0, T]$, $N_{i,t}/S$ is highly likely to remain arbitrarily close to $n_{i,t}$ provided that S is sufficiently large and $N_{i,0}/S = n_{i,0}$ (Appendix S2). Intuitively, for fixed initial densities of both species, larger community sizes correspond to greater population sizes and, consequently, smaller stochastic fluctuations in their densities. However, in contrast to the deterministic model, both species eventually go extinct in finite time in the stochastic model (Appendix S2).

Model parameterization

In our analyses, we use parameter values for our competition model that were estimated for 90 pairwise combinations of 18 annual plant species competing in experimental field plots on serpentine soils in California (Godoy et al., 2014). We used established methods to estimate these parameters (Hart et al., 2018). Briefly, prior to the growing season we sowed focal individuals of each species into a density gradient of competitors. By recording the fecundity of each focal individual near the end of the growing season within a competitive neighborhood, we were able to estimate per-germinant seed production in the absence of competition and the rate of decline in seed production as competitor density increases (Hart et al., 2018). We also measured the germination rate of each species and seed survival rate in the seedbank. See Godoy et al. (2014) for the full details of our empirical and statistical methods.

These experiments were used to parameterize a model of the form

$$n_{i,t+1} = \frac{r_i g_i n_{i,t}}{1 + a_{ii} g_i n_{i,t} + a_{ij} g_j n_{j,t}} + (1 - g_i) s_i n_{i,t}, \quad (3)$$

where $n_{i,t+1}$ is the density of seeds in the soil after seed production but prior to germination, r_i is the per-germinant seed production in the absence of competition, a_{ii} and a_{ij} are the competition coefficients for germinated individuals, g_i is the fraction of germinating seeds, and s_i is seed survival. Because including the seed bank greatly complicates our analysis yet ignoring it would unfairly bias competitive outcomes in the system, we assume that seeds that ultimately germinate do so in the year after they are produced. This implies that higher seed survival rates ultimately increase the

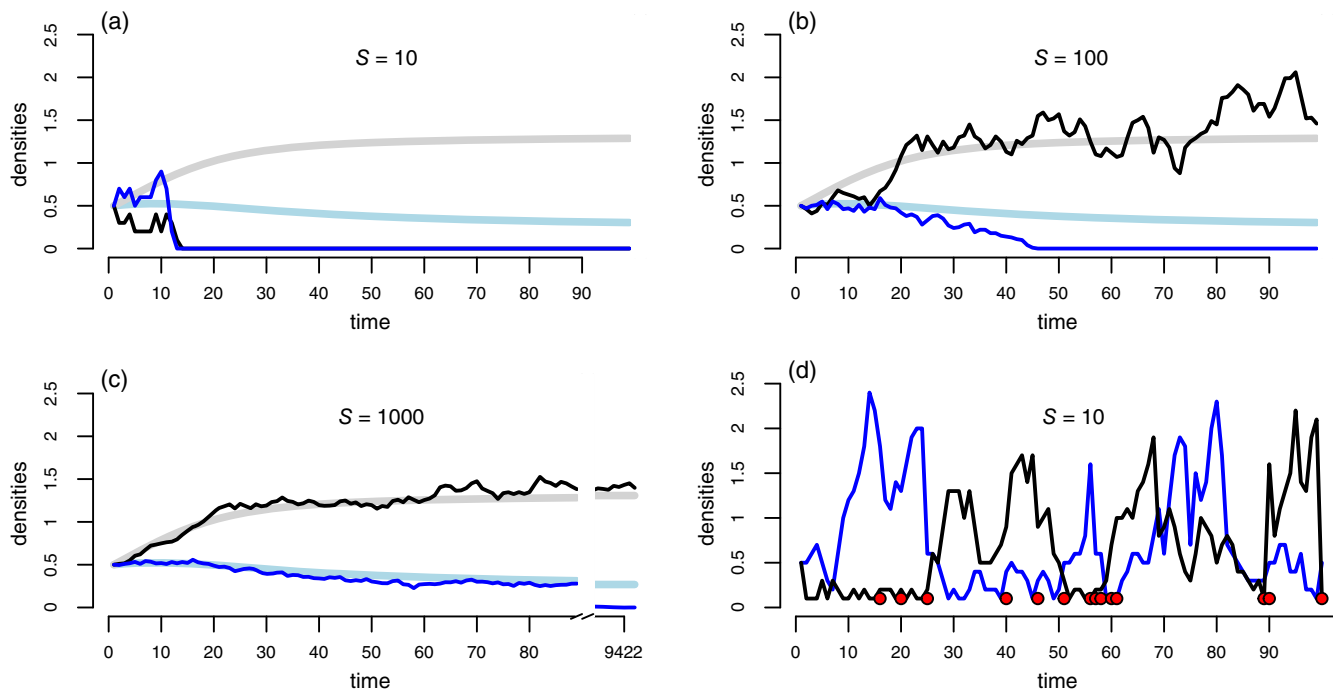


FIGURE 1 (a–c) Deterministic dynamics of species densities $n_{i,t}$ for Equation (1) are shown in light gray and light blue. Sample trajectories of the stochastic dynamics of the species densities $N_{i,t}/S$ for Equation (2) are shown in blue and black. As the community size S increases, there is a closer correspondence between these dynamics, but eventually species go extinct in the stochastic model. (d) Sample simulation of Aldous et al. (1988) algorithm to estimate quasi-stationary distribution and intrinsic coexistence time. A red circle corresponds to times at which one or both species are going extinct; states at these times are replaced by states randomly sampled from the past. The frequency p of these red dots, for sufficiently long runs, determines the intrinsic coexistence time $1/p$.

average annual germination rate, \tilde{g}_i (Appendix S1). After making this adjustment, the model given by Equation (3) is equivalent to the model given by Equation (1) after setting $s = 0$, $\lambda_i = r_i \tilde{g}_i$, and $\alpha_{ij} = \tilde{g}_j a_{ij} / (\lambda_i - 1)$. In the absence of a fluctuating environment (where seeds survival may buffer populations from extinction), this change is expected to have no effect on the duration of species' coexistence.

Metrics of deterministic coexistence

Ultimately, we wish to relate coexistence times to the metrics of deterministic coexistence commonly used in contemporary studies of species coexistence. Here we focus on the invasion growth rate and quantitative definitions of niche and fitness differences, which are themselves derivatives of the invasion growth rate. For Equation (1), the invasion growth rate for species i is given by

$$\mathcal{I}_i = \frac{\lambda_i}{1 + \alpha_{ij} (1/\alpha_{jj}) (\lambda_i - 1)}, \tag{4}$$

where $1/\alpha_{jj}$ is the single-species equilibrium density of species j . In the deterministic model, coexistence occurs

if the invasion growth rates of both species are greater than one. Equivalently, this occurs when the minimum of the invasion growth rates, $\min\{\mathcal{I}_1, \mathcal{I}_2\}$, is greater than one. Conversely, if $\min\{\mathcal{I}_1, \mathcal{I}_2\} < 1$, then one of the species' densities will approach zero over the infinite time horizon.

Each species' invasion growth rate is determined by (1) frequency-dependent processes that provide growth-rate advantages to both species when they are at low relative density and (2) frequency-independent processes that always favor the growth of one species over another, regardless of relative density. These contributions to invasion growth rates have been quantitatively formalized as the niche overlap (ρ) and the average fitness ratio κ_j/κ_i of the two species, respectively (see Chesson [2013] for mathematical details):

$$\rho = \sqrt{\frac{\alpha_{ij} \alpha_{ji}}{\alpha_{jj} \alpha_{ii}}} \text{ and } \frac{\kappa_j}{\kappa_i} = \sqrt{\frac{\alpha_{ii} \alpha_{ij}}{\alpha_{jj} \alpha_{ji}}}. \tag{5}$$

Niche overlap, ρ , decreases as the strength of intraspecific competition increases relative to the strength of interspecific competition. Low niche overlap causes species at high relative density to limit their own growth through intraspecific competition more than they limit the growth

of species at low relative density through interspecific competition, leading to higher invasion growth rates. The fitness ratio κ_j/κ_i increases as one species becomes more sensitive to the total amount of competition in the system. For example, when $\kappa_j/\kappa_i > 1$, species i is more impacted by competition and will be displaced by species j when there is perfect niche overlap. The fitness ratio is used to quantify fitness differences: The greater the magnitude of the log fitness ratio, the greater the fitness difference. Ultimately, both species have positive invasion growth rates when $\rho < \kappa_j/\kappa_i$ and $\rho < \kappa_i/\kappa_j$, demonstrating that deterministic coexistence occurs when niche overlap is less than 1 and small relative to the fitness difference.

If the species deterministically coexist, they will approach a globally stable equilibrium, (n_1^*, n_2^*) . Solving for the densities at which both species' fitness equals 1 in the deterministic model given by Equation (1) gives

$$n_i^* = \frac{\alpha_{ij} - \alpha_{jj}}{\alpha_{ij}\alpha_{ji} - \alpha_{ii}\alpha_{jj}}. \quad (6)$$

Invasion growth rates (\mathcal{I}_i) and equilibrium densities (n_i^*) can also be expressed as follows in terms of the niche overlap and the fitness ratio:

$$\mathcal{I}_i = \frac{\lambda_i}{1 + \rho(\kappa_j/\kappa_i)(\lambda_i - 1)} \quad \text{and} \quad n_i^* = \frac{1}{\alpha_{ii}} \frac{1 - \rho\kappa_j/\kappa_i}{1 - \rho^2}. \quad (7)$$

These expressions will become informative for interpreting the effects of niche overlap and fitness differences on coexistence times.

Estimating intrinsic coexistence times

Unlike in the deterministic model, in the stochastic model, both species always go extinct in finite time. We define the length of time prior to the first species going extinct as the coexistence time. The distribution of coexistence times in the stochastic model will depend on the initial conditions of the system. In studies of single-species persistence, Grimm and Wissel (2004) defined a quantity, the intrinsic mean time to extinction, that provides a common approach for selecting the initial distribution of the population and, thereby, allows for cross-parameter and cross-model comparisons. Here, we extend this work to introduce an analogous concept, the intrinsic coexistence time. The intrinsic coexistence time assumes that competing species have coexisted for a sufficiently long period of time in the past to exhibit relatively stationary population dynamics, that

is, quasi-stationarity. At quasi-stationarity, the distribution of community population sizes is given by the model's quasi-stationary distribution (QSD) (Méléard & Villemonais, 2012). While in this quasi-stationary state, there is a constant probability, call it the persistence probability p , of losing one or both species each time step. We call the mean time $1/p$ to losing at least one of the species the intrinsic coexistence time.

More precisely, the QSD for our stochastic model is a probability distribution $\pi(N_i, N_j)$ on positive population sizes $N_i > 0, N_j > 0$ such that, if the species initially follow this distribution (i.e., $\Pr[N_i(0) = N_i, N_j(0) = N_j] = \pi(N_i, N_j)$ for all $N_i > 0, N_j > 0$), then they remain in this distribution, provided neither goes extinct, i.e.,

$$\begin{aligned} \Pr[N_i(1) = N_i, N_j(1) = N_j | N_i(0) > 0, N_j(0) > 0] \\ = \pi(N_i, N_j) \quad \text{for all } N_i > 0, N_j > 0. \end{aligned}$$

By the law of total probability, the persistence probability p when following the QSD satisfies

$$p = \sum_{N_i > 0, N_j > 0} \Pr[N_i(1) > 0, N_j(1) > 0 | N_i(0) = N_i, N_j(0) = N_j] \pi(N_i, N_j).$$

From a matrix point of view, π corresponds to the dominant left eigenvector of the transition matrix for the stochastic model, and p is the corresponding eigenvalue (Méléard & Villemonais, 2012).

Because the state space for the stochastic model is large even after truncating for rare events, directly solving for the dominant eigenvector is computationally intractable. Fortunately, there is an efficient simulation algorithm for approximating the QSD due to Aldous et al. (1988). This algorithm corresponds to running a modified version of the stochastic model (Figure 1d). Whenever one of the species goes extinct in the simulation, the algorithm replaces this state with a randomly sampled state from the past. More precisely, if $(\tilde{N}_i(1), \tilde{N}_j(1)), \dots, (\tilde{N}_i(t), \tilde{N}_j(t))$ are the states of the modified process until year t , then compute $(\tilde{N}_i(t+1), \tilde{N}_j(t+1))$ according to Equation (2). If $\tilde{N}_i(t+1) = 0$ or $\tilde{N}_j(t+1) = 0$, then replace $(\tilde{N}_i(t+1), \tilde{N}_j(t+1))$ by randomly choosing with equal probability $1/t$ among the prior states $(\tilde{N}_i(1), \tilde{N}_j(1)), \dots, (\tilde{N}_i(t), \tilde{N}_j(t))$. The empirical distribution of $(\tilde{N}_i(1), \tilde{N}_j(1)), \dots, (\tilde{N}_i(t), \tilde{N}_j(t))$ approaches the QSD as $t \rightarrow \infty$. This algorithm converges exponentially fast to the QSD (Benaim & Cloez, 2015). We use this algorithm to

estimate coexistence times in our analyses. In Appendix S8, we also show how this algorithm also applies to a large class of models simultaneously accounting for demographic, environmental stochasticity, and spatial heterogeneity.

Having introduced our model and methods for determining coexistence times, we now describe how we apply these methods to address a series of questions about the relationship between deterministic coexistence and stochastic coexistence times for our empirically parameterized models. We emphasize that, although our models are empirically parameterized and, thus, grounded in real biology, the focus of our analysis is a comparison between the predictions of the empirically parameterized deterministic model and the predictions of the empirically parameterized stochastic model. Understanding the relationship between our predicted coexistence times and observed coexistence times in nature is exceedingly difficult, likely requiring empirical studies over hundreds to thousands of generations in most systems. This is beyond the scope of our current analysis.

How do deterministic coexistence and competitive exclusion relate to coexistence times?

We first explore how deterministic coexistence and deterministic competitive exclusion are related to coexistence times in the presence of demographic stochasticity. We do this through a mixture of analytical and numerical approaches. The analytical approach uses large deviation theory (Faure & Schreiber, 2014) to characterize how intrinsic coexistence times scale with community size S and how this scaling depends on whether the deterministic model predicts coexistence or exclusion. Based on our empirical parameterizations, competition between 82 species pairs is expected to result in deterministic competitive exclusion (i.e., parameter values result in $\min\{\mathcal{I}_1, \mathcal{I}_2\} < 1$), whereas eight species pairs are expected to stably coexist (i.e., parameter values result in $\min\{\mathcal{I}_1, \mathcal{I}_2\} > 1$). The remaining species pairs were excluded from our analyses either because estimates for at least one of the model parameters were missing or because one of the species had an intrinsic fitness λ_i of less than one. For the two groups of species pairs we focus on (i.e., either deterministically coexisting or resulting in deterministic exclusion), we used simulations of 10 million years to compute intrinsic coexistence times C across a range of community sizes S . We describe the relationship between community size and coexistence times for the models of the species pairs in each group.

Do invasion growth rates predict coexistence times?

We used linear regression to determine whether deterministic invasion growth rates as given by Equation (4) influenced intrinsic coexistence times. For this analysis, we calculated intrinsic coexistence times only for the species pairs that were expected to deterministically coexist. To estimate intrinsic coexistence times for these species, we set the community size $S = 0.04$. This size would be empirically justified for the smallest serpentine hummocks in our study landscape (those with a few square meters of suitable habitat), which might contain as few as 20 individuals of the subdominant species (based on germinable densities projected from Gilbert & Levine, 2013). The species that compose the focal pairs in this analysis tend to be more common and are often found on larger hummocks, but this small habitat size allows us to evaluate the dynamics of systems where the effects of demographic stochasticity might be substantial. For each species pair we used simulations of 10^7 years to estimate the coexistence times predicted by the models. For seven of the species pairs predicted to coexist by the deterministic models, we were able to generate good estimates of intrinsic coexistence times predicted by the stochastic models because there were multiple extinction events in simulations of this length. For one species pair predicted to coexist by the deterministic model, there were no extinction events in the numerical simulations. Because we only had a lower bound of $>10^7$ years for the mean intrinsic coexistence time for this pair, it was excluded from our analysis.

Coexistence time was the dependent variable in our regression, and the log of $\min\{\mathcal{I}_1, \mathcal{I}_2\}$ was the independent variable. We used this independent variable because the species with the lower invasion growth rate is, all else being equal, more likely to go extinct first. To examine the robustness of our conclusions, we repeated our analysis for 1000 randomly drawn parameter values with community size $S = 10$. These random draws were performed on uniform distributions with $[1.1, 2]$ for the λ_i values, $[0.2, 0.5]$ for the α_{ii} values, and $\alpha_{ii} \times [0, 1]$ for the α_{ij} values (to ensure deterministic coexistence).

Do equilibrium population sizes predict coexistence times?

Invasion growth rates were a less than perfect predictor of coexistence times for the stochastic models (see *Results*). Therefore, we also used linear regression to explore the relationship between the equilibrium population sizes of the coexisting species and the intrinsic coexistence time. For this analysis, we used the same methods as described previously, but with the log of the

minimum of the two equilibrium population sizes, that is, $\min\{Sn_1^*, Sn_2^*\}$, as the independent variable in the linear regression. To examine the robustness of our conclusions, we repeated our analysis for 1000 randomly drawn parameter values with community size $S = 10$ as described earlier in the section *Do invasion growth rates predict coexistence times?*

Do greater niche overlap or greater fitness differences always reduce coexistence times?

According to deterministic theory, greater niche overlap and greater fitness differences both have a negative impact on coexistence (Adler et al., 2007; Chesson, 2000a; May, 1975). It is important, therefore, to understand whether these negative effects on deterministic coexistence also have predictably negative effects on the duration of species' coexistence. We investigate these relationships by calculating coexistence times as a function of niche overlap and fitness differences.

Our goal is to assess the independent effects of these determinants of coexistence on coexistence times. However, because niche overlap and fitness differences are both functions of the same parameters given by Equation (5), they are not quantitatively (or biologically) inherently independent (see also Song et al., 2019). This precludes using the natural variation in niche overlap and fitness differences observed across our focal species pairs for our analyses. Therefore, to achieve our goal, we manipulated niche overlap and fitness differences separately for each species pair. Specifically, to assess the effects of niche overlap on model-predicted coexistence times, we multiplied the interspecific competition coefficients α_{12}, α_{21} for each coexisting species pair by a common, fixed factor. This manipulation allows the niche overlap to vary while keeping the fitness ratio constant according to Equation (5). Mechanistically, this may be interpreted as altering the degree to which the two species use the same resources. Similarly, to assess the effects of the fitness difference independent of any change in niche overlap, we multiplied the competition coefficients α_{22}, α_{21} within each coexisting pair by a fixed factor. This manipulation reduces the sensitivity of Species 2 to competition, increasing its competitive ability according to Equation (5), while having no effect on niche overlap. Mechanistically, this may be interpreted as increasing the efficiency with which Species 2 uses shared resources. Based on these manipulations, for each coexisting species pair we explored the relationship between niche overlap and coexistence times, and the relationship between fitness differences and coexistence times.

We interpret these relationships via the effects of niche overlap and fitness differences on invasion growth rates and equilibrium population sizes as per Equation (7).

Do stochastic competitive dynamics influence coexistence times?

Competition may also influence coexistence times because the stochastic population dynamics of two species are coupled. For example, a stochastic increase in the population size of one species might be expected to result in a concomitant decrease in the population size of its competitor, increasing its risk of extinction. The effect on the coexistence times of this dynamic coupling of the two species is not accounted for by static metrics of coexistence, such as invasion growth rates and equilibrium population sizes. To identify whether this dynamic coupling plays an important role in determining coexistence times, we built a simplified model, which we call the demographically uncoupled model, incorporating the effects of competition on invasion growth rates and equilibrium population sizes but excluding the coupling of the stochastic fluctuations in the population sizes of the competitors. In our demographically uncoupled model, each species has a low-density growth rate, λ_i , and carrying capacity, $1/\alpha_{ii}$, that matches its invasion growth rate \mathcal{I}_i and equilibrium density n_i^* , respectively, from the deterministic two-species model. The update rule for this simplified model is

$$N_{i,t+1} = \text{Pois} \left(\frac{\mathcal{I}_i N_{i,t}}{1 + \frac{N_{i,t}}{Sn_i^*} (\mathcal{I}_i - 1)} \right) \text{ for } i = 1, 2. \quad (8)$$

Importantly, note that the change in the population size of N_i does not depend on N_j . Therefore, in the demographically uncoupled model, the effects of interspecific competition on invasion growth rates and equilibrium population sizes are retained, but the species are dynamically uncoupled. By quantifying the relationship between coexistence times calculated for the full (Equation 2) and demographically uncoupled (Equation 8) models, we can determine whether the combined effects of competition on invasion growth rates and equilibrium population sizes are sufficient to predict coexistence times or whether competitive dynamics arising from the coupled stochastic dynamics of the species are also important.

In Appendix S7, we show the coexistence times for the demographically uncoupled model reduce to calculating the intrinsic persistence times P_i of each species in this demographically uncoupled model. These persistence times P_i are calculated independently for each species using the

Aldous et al. (1988) algorithm. Then the coexistence time $C_{\text{uncoupled}}$ of the demographically uncoupled model equals

$$C_{\text{uncoupled}} = \frac{P_1 P_2}{P_1 + P_2 - 1}. \tag{9}$$

When the intrinsic persistence times P_i for each species are sufficiently long (e.g., 10 years or more), the coexistence time predicted by this demographically uncoupled model is approximately one half of the harmonic mean of the intrinsic persistence times for the two species: $C_{\text{uncoupled}} \approx \frac{1}{2} \times \frac{1}{\frac{1}{2P_1} + \frac{1}{2P_2}}$. Because the harmonic mean is dominated by the minimum of its arguments, the species with the shorter persistence time has a greater influence on the coexistence time.

We calculated intrinsic coexistence times for the full and demographically uncoupled models for the eight species pairs expected to deterministically coexist. We then used linear regression to determine the extent to which the demographically uncoupled model explains coexistence times in the full model. To examine the robustness of our conclusions, we repeated our analysis for 1000 randomly drawn parameter values with community size $S = 10$, as described in the earlier section *Do invasion growth rates predict coexistence times?*

RESULTS

How do deterministic coexistence and competitive exclusion relate to coexistence times?

The duration of species' coexistence tended to be several orders of magnitude larger for species expected to

deterministically coexist than for those pairs where deterministic exclusion is expected (Figure 2). Indeed, our numerical results demonstrate that species expected to deterministically coexist will tend to do so for decades to millennia, even when community sizes are small. By contrast, our numerical simulations suggest that the vast majority of the 82 species pairs for which deterministic exclusion is expected will tend to coexist for less than 5 years (Figure 2a).

Our analytical and numerical results reveal a fundamental dichotomy between deterministic coexistence and exclusion in terms of how coexistence times for the models scale with increasing community size. In particular, if deterministic exclusion is predicted ($\min\{\mathcal{I}_1, \mathcal{I}_2\} < 1$), then our mathematical analysis (Appendix S3) implies that the intrinsic coexistence times are bounded above by $1/(1 - \min\{\mathcal{I}_1, \mathcal{I}_2\})$ and do not increase significantly with community size. This analytical result is consistent with our numerical results for the 82 models associated with the species pairs for which deterministic exclusion was predicted (Figure 2a). By contrast, if the two species are predicted to deterministically coexist ($\min\{\mathcal{I}_1, \mathcal{I}_2\} > 1$), then our analysis implies that intrinsic coexistence times scale exponentially with the community size S . This means that small increases in community size lead to very large increases in the duration of species' coexistence. This analytical result is also consistent with our simulations of the eight species pairs for which deterministic exclusion was predicted (Figure 2b).

For species pairs predicted to deterministically coexist, the mathematical analysis only ensures there is a constant $\star > 0$ such that the intrinsic coexistence time is proportional to $e^{\star S}$. There is no simple formula for \star because it will depend on many details of the individual-based model (see proofs in Faure and

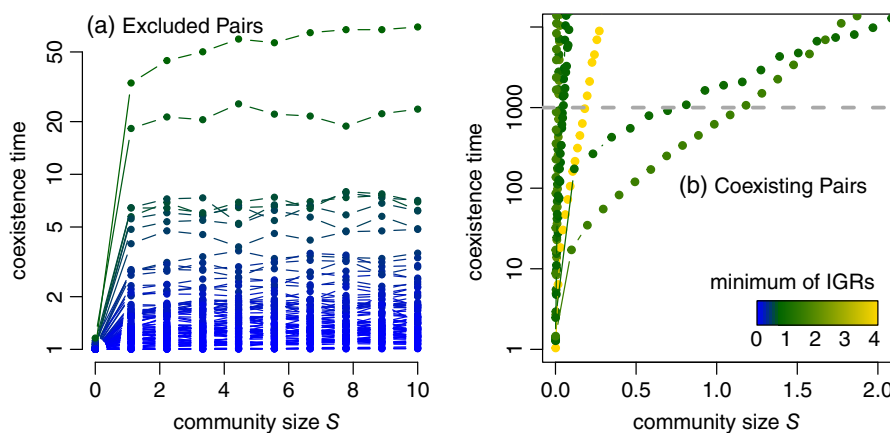


FIGURE 2 Coexistence times as a function of community size S for (a) deterministically excluded species pairs and (b) deterministically coexisting pairs. For each viable species pair, the invasion growth rates $\mathcal{I}_1, \mathcal{I}_2$ were computed for the associated deterministic model. For each species pair, the coexistence time of the associated individual-based model were calculated for a range of community sizes S . Each of these species pair curves are colored by their $\min\{\mathcal{I}_1, \mathcal{I}_2\}$ value.

Schreiber 2014). To understand some of these dependencies, we explored how coexistence times depended on key coexistence metrics from the deterministic model for a fixed community size $S = 0.04$ (a small hummock; see methods), the results of which we describe next.

Do invasion growth rates predict coexistence times?

Across seven pairs of deterministically coexisting competitors, we found a positive correlation between the minimum invasion growth rate, $\min\{\mathcal{I}_1, \mathcal{I}_2\}$, and intrinsic coexistence times, as determined from the QSDs of the empirically parameterized model (adjusted $R^2 = 0.6038$, $p = 0.0244$; Figure 3a). Intuitively, the species with the lower invasion growth rate tends to go extinct first, so it is the lower of the two invasion growth rates that predicts coexistence times. However, there was substantial variation in coexistence times unexplained by invasion growth rates, such that similar invasion growth rates resulted in a difference of several orders of magnitude in coexistence times (Figure 3a). Random sampling of parameter space produced similar results (Appendix S4: Figure S1A).

Do equilibrium population sizes predict coexistence times?

We also found a positive correlation between the minimum of the equilibrium population sizes, $\min\{Sn_1^*, Sn_2^*\}$, and coexistence times (adjusted $R^2 = 0.3386$; $p = 0.09964$; Figure 3b). Because equilibrium population sizes did a worse job of explaining the variation in the coexistence

times than the invasion growth rates, similar equilibrium population sizes also resulted in several orders of magnitude of difference in coexistence times. Random sampling of parameter space produced similar results (Appendix S4: Figure S1B).

Does greater niche overlap or greater fitness differences always reduce coexistence times?

Greater niche overlap and greater fitness differences did not always reduce coexistence times, as deterministic theory would predict (Figures 4 and 5). For five out of seven species pairs, both greater niche overlap and greater fitness differences reduced the coexistence times predicted by the models (Figures 4a,d; Appendix S6). For these pairs, the relationships were nonlinear but negatively monotonic. However, for the remaining two species pairs, the effects of greater niche overlap and fitness differences on coexistence times were complex (Figures 5a,d; Appendix S6). For these species pairs, increasing niche overlap consistently reduced coexistence times, but in a highly nonlinear fashion (Figure 5a). By contrast, increasing fitness differences had both positive and negative effects on coexistence times for the models associated with these species pairs, depending on the magnitude of the change in the fitness difference (Figure 5d). Importantly, for the species pairs where these complex effects emerged, the competitively inferior species (as determined by the fitness ratio in Equation 5), had the higher (i.e., not the minimum) equilibrium population size.

When the inferior competitor has the higher equilibrium population size, complex effects on coexistence times emerge via the influence of niche overlap and fitness differences on the minimum equilibrium population sizes and the minimum invasion growth rates (Equation 7). For example, as fitness differences increase, the superior competitor's equilibrium population size also increases (Equation 7, Appendix S6, Figure 5e,f). Consequently, when the superior competitor has a lower equilibrium population size, increasing the fitness difference increases the minimum of the equilibrium population sizes, which increases coexistence times (Figure 5d). Increasing niche overlap can also increase the equilibrium population size of the superior competitor, but it always decreases its invasion growth rate (Equation 7, Appendix S6, Figure 5b,c). When the superior competitor has a lower equilibrium abundance, these opposing trends result in coexistence times that decrease in a highly nonlinear manner with increasing niche overlap. A general result is that coexistence times decrease when both the minimum invasion growth rate and the minimum equilibrium population size

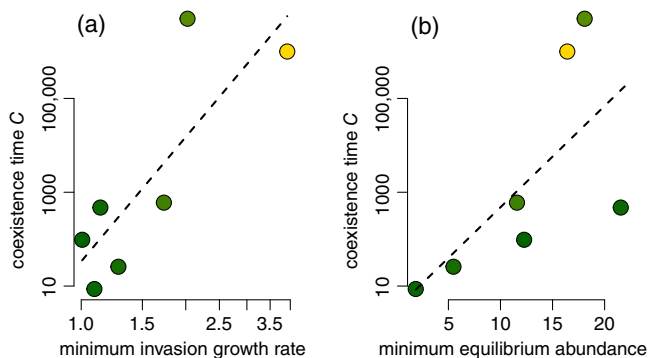


FIGURE 3 Predicting coexistence times using the (a) minimum, $\min\{\mathcal{I}_1, \mathcal{I}_2\}$, of invasion growth rates and the (b) minimum, $\min\{Sn_1^*, Sn_2^*\}$, of equilibrium population sizes. Both panels plotted on a log₁₀ scale. Each of these species pair curves are colored by their $\min\{\mathcal{I}_1, \mathcal{I}_2\}$ value as in Figure 2.

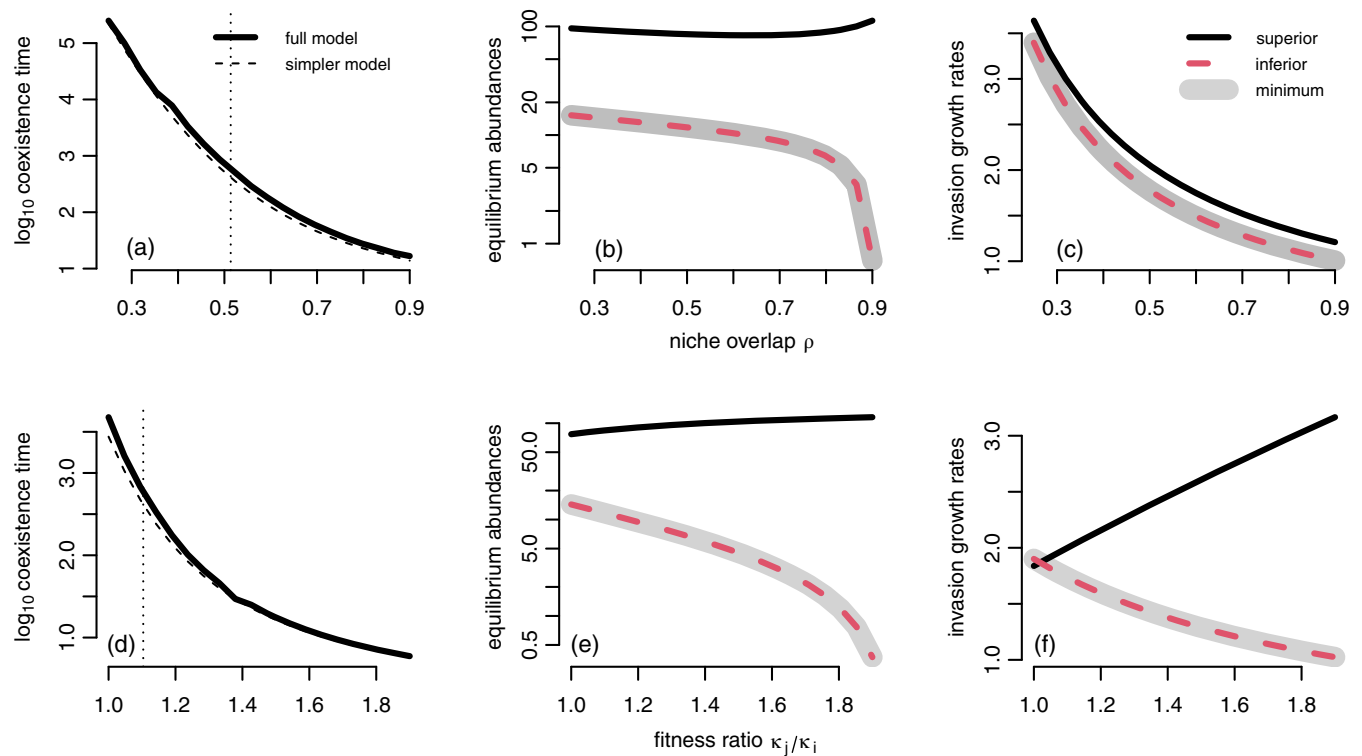


FIGURE 4 Effects of niche overlap and fitness ratios on coexistence times for a pair of competing species where the fitness inferior has the lower equilibrium abundance. In (a) and (d), \log_{10} coexistence times and the demographically uncoupled model approximations (dashed lines) are plotted. The vertical dotted line corresponds to the base empirical value. In (b) and (e), the deterministic equilibrium population sizes Sn_i^* are plotted, with the gray shading indicating the minimum of the two population sizes. In (c) and (f), the deterministic invasion growth rates of the deterministic, mean field model are plotted for the fitness inferior (dashed red) and fitness superior (solid black), with gray shading indicating minimum of two invasion growth rates.

decline with increasing niche overlap and increasing fitness differences (Figures 4 and 5).

Do stochastic competitive dynamics influence coexistence times?

To isolate the influence of coupled stochastic competitive dynamics on coexistence times, we compared coexistence times calculated from the full model given by Equation (2), including these coupled dynamics, with coexistence times calculated from the demographically uncoupled model given by Equation (8), excluding these coupled dynamics. As described in *Models and methods*, the demographically uncoupled model is two uncoupled, individual-based single-species models whose low-density growth rate and intraspecific competition coefficients equal T_i and $1/n_i^*$, respectively. For the seven species pairs predicted to coexist in the deterministic model, the demographically uncoupled model incorporating competition only through its effects on invasion growth rates and equilibrium population sizes did an exceptional job in predicting the actual coexistence

time ($\log C = 1.031 \log C_{\text{uncoupled}}$ with $R^2 = 0.9977$ and $p < 10^{-8}$, Figure 6). Random sampling of parameter space produced similar results (Appendix S4: Figure S1C).

DISCUSSION

We introduced a new metric, the intrinsic coexistence time, that characterizes the risk of species loss due to demographic stochasticity over any time horizon. This metric complements more traditional coexistence metrics based on invasion growth rates and can be computed for many existing data-based models by following a two-step procedure. First, extend the deterministic model to account for demographic stochasticity. Informing such a stochastic model with demographic data obtained in a field context is not much more difficult than parameterizing deterministic models. Aside from fecundity, most transition rates (i.e., survival, growth, dispersal probabilities) of the deterministic model directly transfer to the stochastic model. The fecundity distribution can be estimated from the raw data used to calculate mean fecundity in deterministic models, or one can assume

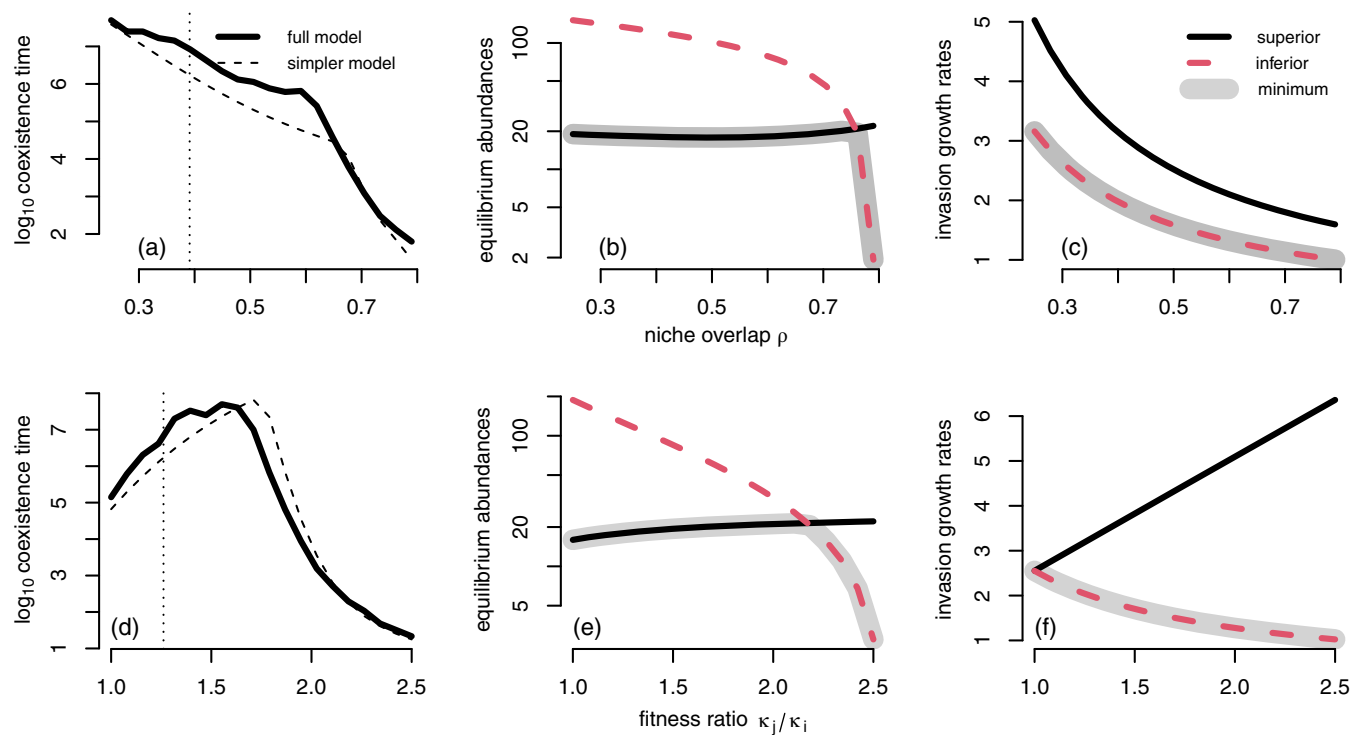


FIGURE 5 Effects of niche overlap and fitness ratios on coexistence times for a pair of competing species, where the fitness superior has the lower equilibrium abundance. In (a) and (d), \log_{10} coexistence times and the demographically uncoupled model approximations (dashed lines) are plotted. The vertical dotted line corresponds to the base empirical value. In (b) and (e), the deterministic equilibrium population sizes Sn_i^* are plotted, with the gray shading indicating the minimum of the two population sizes. In (c) and (f), the deterministic invasion growth rates of the deterministic, mean field models are plotted for the fitness inferior (dashed red) and fitness superior (solid black), with gray shading indicating the minimum of two invasion growth rates.

that fecundity is Poisson distributed with the calculated mean. Second, estimate each intrinsic coexistence time by a single simulation using the algorithm of Aldous et al. (1988) (see Appendix S8 for extensions to a broad class of models). Using this two-step approach, we evaluated how well metrics from deterministic theory predicted intrinsic coexistence times for 18 species of California annuals.

Consistent with the deterministic theory, we found that invasion growth rates were, in general, good predictors of intrinsic coexistence times in models explicitly accounting for demographic stochasticity, thereby reinforcing the general usefulness of these invasion growth rates in theoretical and empirical studies of species coexistence. Their usefulness stems from two conclusions in our study. First, when both species' invasion growth rates are positive (i.e., deterministic coexistence is predicted), intrinsic coexistence times increase exponentially with community size. Thus, for a given minimum invasion growth rate, doubling community sizes quadruples coexistence times. Strikingly, for the eight species pairs of serpentine annuals predicted to deterministically coexist, community sizes corresponding to only 10 individuals of the less common species were still sufficient

to ensure predicted coexistence times of greater than 1000 years (Figure 2)—a time frame well beyond most empirically relevant scenarios and much greater than that considered in most conservation studies (Meine, 1999). Second, for a given community size, we found that the minimum invasion growth rates ($\min\{\mathcal{I}_1, \mathcal{I}_2\}$) explained 60% of the variation in the coexistence times of the serpentine annuals predicted to deterministically coexist (Figure 3) and 82% of the variation of coexistence times for a random sampling of parameter space (Appendix S4: Figure S1A). The two conclusions are consistent with recent work on coexistence times for the lottery model, which simultaneously account for environmental stochasticity and demographic stochasticity (Ellner et al., 2020; Pande, Fung, Chisholm, & Shnerb, 2020), the main differences being that invasion growth rates for the lottery model in the presence of temporal environmental stochasticity correspond to the geometric mean of fitness across time, with coexistence times increasing as a power law, rather than exponentially, with community size (Ellner et al., 2020).

Despite explaining a significant amount of variation, invasion growth rates are not perfect predictors of coexistence times in models accounting for demographic

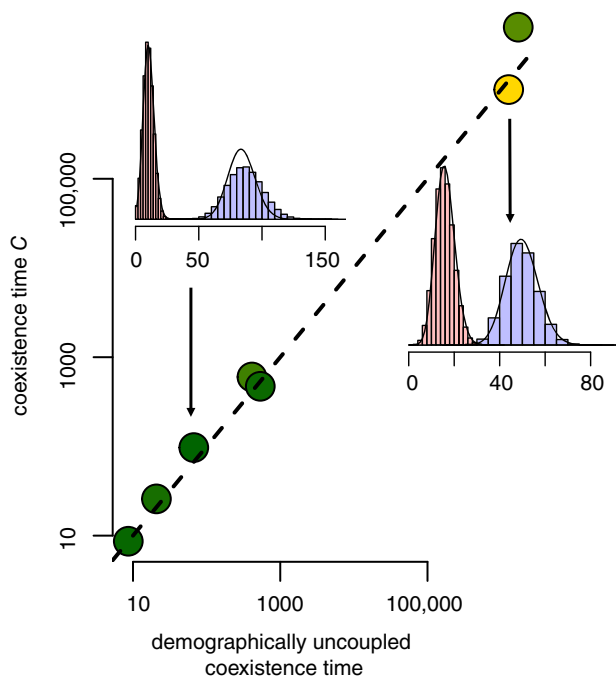


FIGURE 6 Predicting coexistence times with demographically uncoupled model given by Equation (8). Dashed line is the 1-to-1 line. The main panel shows how the demographically uncoupled models incorporating only invasion growth rates and equilibrium population sizes predict coexistence times for the full, demographically coupled models. Insets: Bar plots show simulation-based estimates of coupled, competitive model’s quasi-stationary distributions; black curves are analytical estimates from demographically uncoupled model. Each of these species pair curves are colored by their $\min\{\mathcal{I}_1, \mathcal{I}_2\}$ value, as in Figure 2.

stochasticity. In particular, predicted coexistence times varied by an order of magnitude for species pairs of serpentine annuals with similar values of $\min\{\mathcal{I}_1, \mathcal{I}_2\}$ (Figure 3). Thus, although higher invasion growth rates tend to be associated with longer coexistence times, and these coexistence times tend to be exceedingly long, the capacity of invasion growth rates to accurately predict coexist times should still be viewed with some caution. Pande, Fung, Chisholm, and Shnerb (2020) raised a similar concern but for different reasons when studying Chesson’s lottery model. They demonstrated that increases in environmental variability could simultaneously lead to an increase in invasion growth rates and shorter coexistence times. Our study finds that some disassociation between invasion growth rates and coexistence times can occur even without temporal fluctuations in the fitness of the rare species.

Beyond invasion growth rates, we found that population sizes far from the invasion boundary (i.e., equilibrium population sizes) were also important determinants

of coexistence times. This finding is consistent with single-species models of extinction risk, where both low-density growth rate and equilibrium population size determine risk of extinction (e.g., Lande et al., 2003). However, in and of themselves, equilibrium population sizes were poorer predictors of coexistence times than invasion growth rates. The minimum equilibrium population sizes explained 26% less of variation than $\min\{\mathcal{I}_1, \mathcal{I}_2\}$ in the empirical species pairs (Figure 3) and 12% less of the variation of the coexistence times in the random sampling of parameter space (Appendix S4: Figure S1B).

When combined, the invasion growth rates and the equilibrium population sizes are such good determinants of coexistence times that just these two quantities can explain over 99% of the variation of the predicted coexistence times (Figure 6, Appendix S4: Figure S1C). Notably, this predictive capacity emerges from a demographically uncoupled model parameterized only with the invasion growth rates and equilibrium population sizes, while excluding the effects of the coupled stochastic dynamics of interspecific competition. The success of the demographically uncoupled model emphasizes that coexistence times in the presence of demographic stochasticity can be predicted using minimal information gleaned directly from an appropriate deterministic model. Whether this tractable approach applies to stochastic models with more species or different nonlinearities in the per-capita growth rates represents an important direction for future research.

Since the predicted coexistence time for the demographically uncoupled model is approximately the harmonic mean of the persistence time for each of the (recalibrated) single-species models, one can use it to gain additional insights into coexistence times for competing species. For example, if both competitors have large invasion growth rates (i.e., $\mathcal{I}_i \gg 1$), then each competitor’s persistence time is approximately equal to its exponentiated equilibrium population abundance, $\exp(n_i^*S)$ (Appendix S7), in which case, for a given community size $n_1^*S + n_2^*S = N$, the predicted coexistence time is maximized when both species have equal equilibrium abundances (Appendix S7). For a more diverse community, this is equivalent to when Shannon’s diversity index is maximized. However, if one species has a significantly lower invasion growth rate than its competitor, then maximizing the coexistence time requires that this species be overrepresented in the community (i.e., the community would have a lower Shannon diversity index) (Appendix S7).

Because invasion growth rates are imperfect predictors of coexistence times, coexistence metrics derived solely from invasion growth rates are also imperfect

predictors of coexistence times. These coexistence metrics include modern quantitative definitions of niche overlap and fitness differences (Chesson, 2000a, 2013, 2018; Godoy & Levine, 2014). Because niche overlap and fitness differences can have opposing effects on the relative population sizes of competitors, coexistence times do not necessarily decrease with increasing niche overlap or increasing fitness differences, as would be expected based only on deterministic theory (Figure 5a,d). For models of two of the annual, serpentine species pairs, this unexpected outcome occurred when the inferior competitor had the higher equilibrium population size. Operationally, one can simply check whether this is the case before interpreting the effects of niche overlap and fitness differences as they are commonly derived and applied (Adler et al., 2007, 2010; Gilbert & Levine, 2013; Chesson, 2000a, 2013). According to Equation (7), the competitive inferior has the larger equilibrium abundance when it is insensitive to intraspecific competition (i.e., has a large carrying capacity $1/\alpha_{jj}$) but highly sensitive to interspecific competition relative to the competitively superior species. This observation provides an indirect means of evaluating whether this inverted relationship is likely to occur in a given empirical system.

Our conclusions are based on an analysis of an annual plant competition model, which is a specific case of a large and more general class of models that can be reexpressed in Lotka–Volterra form. This class of models has been the focus of much of the theoretical and empirical work deriving and applying quantitative definitions of niche and fitness differences to studies of species coexistence (Chesson, 1990, 2000a, 2013; Godoy & Levine, 2014; MacArthur & Levins, 1967). Given that niche and fitness differences in these models are derived from invasion criteria, we expect our conclusions about the effects of niche and fitness differences on coexistence times to apply across this class of models. More generally, equilibrium densities of competitors are likely to be important quantities in all models of competition regardless of their complexity but were not often taken into account in recent assessments of coexistence (Carroll et al., 2011; Chesson, 2013; Spaak & De Laender, 2020). For example, Spaak & De Laender (2020) provided a general method for defining niche and fitness differences using per-capita growth rates evaluated at densities where at least one species is absent. However, these metrics do not take into account, and are unlikely to correlate with, coexistence equilibrium densities. Therefore, coexistence times are unlikely to map intuitively to niche and fitness differences even for these more general definitions. In sum, our work suggests

that any coexistence metric not explicitly taking into account densities at which the species coexist deterministically may not predict coexistence times correctly.

For models considered here, all conspecific individuals have the same invasion growth rate across time—there is no variation in the invasion growth rates among subpopulations of individuals. However, phenotypic, spatial, and temporal variation in vital rates can impact competitive outcomes (Chesson, 1994, 2000b; Hart et al., 2016; Levin, 1974; Schreiber et al., 2011; Stump et al., 2021; Vasseur et al., 2011; Warner & Chesson, 1985). Variation in vital rates often results in variation in invasion growth rates among subpopulations of individuals within generations (spatial or phenotypic variation) or between generations (temporal variation). In these situations, the distribution of these invasion growth rates, not only the mean, can play a role in coexistence times. Understanding their role and how coexistence times depend on the nature of the variation (phenotypic vs. spatial vs. temporal) is a major challenge for future work. Because our methodology for computing intrinsic coexistence is applicable to models accounting for this variation (see details in Appendix S8), it may provide a unified approach to tackling this challenge.

Some progress has been made on understanding the role of temporal variation on coexistence times (Ellner et al., 2020; Pande et al., 2020; Pande, Fung, Chisholm, & Shnerb, 2020). For example, using the methods presented here, Ellner et al. (2020) found that the mean invasion growth rates for the lottery model (Warner & Chesson, 1985) were strongly correlated with intrinsic coexistence times. However, Pande et al. (2020a, 2020b) showed that ignoring the temporal variation in invasion growth rates could result in incorrect inferences about coexistence times. For example, greater environmental variation can increase mean invasion growth rates via the storage effect (Chesson, 1994), but they can also lead to shorter coexistence times via the storage effect. Similar conclusions may also apply to within-generation sources of variation. For example, increasing within-generation variation in fecundity decreases the persistence times of single species (Melbourne & Hastings, 2008). We anticipate similar effects on coexistence times in our models.

The complexities we have identified in our study highlight the need to understand deterministic coexistence from both equilibrium-based and invasion-based approaches. Recent relevant equilibrium-based theories include Saavedra et al.'s (2017) structural stability of feasible equilibria and Barabás et al.'s (2014) sensitivity analysis of stable equilibria. Both of these approaches

provide insights into the range of demographic parameter values for which the equilibrium abundances remain positive, that is, feasible. Moreover, Saavedra et al. (2017) developed feasibility metrics analogous to stabilizing niche differences and fitness differences. These developments raise the promising possibility of developing more informative, integrative metrics for species coexistence times.

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CONFLICT OF INTEREST

The authors declare no conflict of interest.

DATA AVAILABILITY STATEMENT

Code (Schreiber, 2022) is available on Figshare at <https://doi.org/10.6084/m9.figshare.17125598.v1>.

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REFERENCES

- Adler, P., S. Ellner, and J. Levine. 2010. "Coexistence of Perennial Plants: An Embarrassment of Niches." *Ecology Letters* 13: 1019–29.
- Adler, P., J. HilleRisLambers, and J. Levine. 2007. "A Niche for Neutrality." *Ecology Letters* 10: 95–104.
- Adler, P. B., and J. M. Drake. 2008. "Environmental Variation, Stochastic Extinction and Competitive Coexistence." *The American Naturalist* 172: E186–95.
- Aldous, D., B. Flannery, and J. Palacios. 1988. "Two Applications of Urn Processes the Fringe Analysis of Search Trees and the Simulation of Quasi-Stationary Distributions of Markov Chains." *Probability in the Engineering and Informational Sciences* 2: 293–307.
- Barabás, G., L. Pásztor, G. Meszéna, and A. Ostling. 2014. "Sensitivity Analysis of Coexistence in Ecological Communities: Theory and Application." *Ecology Letters* 17(12): 1479–94.
- Benaïm, M., and B. Cloez. 2015. "A Stochastic Approximation Approach to Quasi-Stationary Distributions on Finite Spaces." *Electronic Communications in Probability* 20: 1–14.
- Beverton, R. J. H., and S. J. Holt. 1957. *On the Dynamics of Exploited Fish Populations, volume 2(19) of Fisheries Investigation Series*. London: Ministry of Agriculture, Fisheries and Food.
- Boyce, M. 1992. "Population Viability Analysis." *Annual Review of Ecology and Systematics* 23: 481–97.
- Carroll, I., B. Cardinale, and R. Nisbet. 2011. "Niche and Fitness Differences Relate the Maintenance of Diversity to Ecosystem Function." *Ecology* 92: 1157–65.
- Chesson, P. 1990. "Macarthur's Consumer-Resource Model." *Theoretical Population Biology* 37: 26–38.
- Chesson, P. 1994. "Multispecies Competition in Variable Environments." *Theoretical Population Biology* 45: 227–76.
- Chesson, P. 2000a. "Mechanisms of Maintenance of Species Diversity." *Annual Review of Ecology and Systematics* 31: 343–66.
- Chesson, P. 2000b. "General Theory of Competitive Coexistence in Spatially-Varying Environments." *Theoretical Population Biology* 58: 211–37.
- Chesson, P. 2013. "Species Competition and Predation." In *Ecological Systems* 223–56. New York: Springer.
- Chesson, P. 2018. "Updates on Mechanisms of Maintenance of Species Diversity." *Journal of Ecology* 106: 1773–94.
- Chu, C., and P. Adler. 2015. "Large Niche Differences Emerge at the Recruitment Stage to Stabilize Grassland Coexistence." *Ecological Monographs* 85: 373–92.
- Cushing, J., S. Leverage, N. Chitnis, and S. Henson. 2004. "Some Discrete Competition Models and the Competitive Exclusion Principle." *Journal of Difference Equations and Applications* 10: 1139–51.
- Ellner, S., R. Snyder, and P. Adler. 2016. "How to Quantify the Temporal Storage Effect Using Simulations Instead of Math." *Ecology Letters* 19: 1333–42.
- Ellner, S., R. Snyder, P. Adler, G. Hooker, and S. Schreiber. 2020. "Technical Comment on Pande et al., (2020): Why Invasion Analysis Is Important for Understanding Coexistence." *Ecology Letters* 23(11): 1721–4.
- Ellner, S. P., R. E. Snyder, P. B. Adler, and G. Hooker. 2019. "An Expanded Modern Coexistence Theory for Empirical Applications." *Ecology Letters* 22: 3–18.
- Ewens, W. J. 2012. *Mathematical Population Genetics I. Theoretical Introduction*, Vol. 27. New York, NY: Springer Science & Business Media.
- Faure, M., and S. J. Schreiber. 2014. "Quasi-Stationary Distributions for Randomly Perturbed Dynamical Systems." *Annals of Applied Probability* 24: 553–98.
- Gabel, A., B. Meerson, and S. Redner. 2013. "Survival of the Scarcer." *Physical Review E, Statistical, Nonlinear, and Soft Matter Physics* 87: 010101.
- Gilbert, B., and J. Levine. 2013. "Plant Invasions and Extinction Debts." *Proceedings of the National Academy of Sciences* 110: 1744–9.
- Godoy, O., N. Kraft, and J. Levine. 2014. "Phylogenetic Relatedness and the Determinants of Competitive Outcomes." *Ecology Letters* 17: 836–44.
- Godoy, O., and J. M. Levine. 2014. "Phenology Effects on Invasion Success: Insights from Coupling Field Experiments to Coexistence Theory." *Ecology* 95: 726–36.
- Gómez-Corral, A., and M. López García. 2012. "Extinction Times and Size of the Surviving Species in a Two-Species Competition Process." *Journal of Mathematical Biology* 64: 255–89.

- Grainger, T. N., J. M. Levine, and B. Gilbert. 2019. "The Invasion Criterion: A Common Currency for Ecological Research." *Trends in Ecology & Evolution* 34: 925–35.
- Grimm, V., and C. Wissel. 2004. "The Intrinsic Mean Time to Extinction: A Unifying Approach to Analyzing Persistence and Viability of Populations." *Oikos* 105: 501–11.
- Hart, S., S. Schreiber, and J. Levine. 2016. "How Variation between Individuals Affects Species Coexistence." *Ecology Letters* 19: 825–38.
- Hart, S. P., R. P. Freckleton, and J. M. Levine. 2018. "How to Quantify Competitive Ability." *Journal of Ecology* 106: 1902–9.
- Hofbauer, J., and K. Sigmund. 1998. *Evolutionary Games and Population Dynamics*. Cambridge, UK: Cambridge University Press.
- Hubbell, S. P. 2001. *The Unified Neutral Theory of Biodiversity and Biogeography*, vol. 32 of *Monographs in Population Biology*. Princeton, NJ: Princeton University Press.
- Hubbell, S. P., and R. B. Foster. 1986. *Biology, Chance and History and the Structure of Tropical Rainforest Tree Communities* 314–29. New York: Harper and Row.
- Hutchinson, G. 1961. "The Paradox of the Plankton." *The American Naturalist* 95: 137–45.
- Jagers, P. 2010. "A Plea for Stochastic Population Dynamics." *Journal of Mathematical Biology* 60: 761–4.
- Kramer, A. M., and J. M. Drake. 2014. "Time to Competitive Exclusion." *Ecosphere* 5: 1–16.
- Lande, R., S. Engen, and B.-E. Saether. 2003. *Stochastic Population Dynamics in Ecology and Conservation*. Oxford Series in Ecology and Evolution. Oxford, UK: Oxford University Press.
- Leslie, P. H., and J. C. Gower. 1958. "The Properties of a Stochastic Model for Two Competing Species." *Biometrika* 45: 316–30.
- Letten, A., P. Ke, and T. Fukami. 2017. "Linking Modern Coexistence Theory and Contemporary Niche Theory." *Ecological Monographs* 87: 161–77.
- Levin, S. 1974. "Dispersion and Population Interactions." *The American Naturalist* 108: 207–28.
- Levine, J. M., and J. H. R. Lambers. 2009. "The Importance of Niches for the Maintenance of Species Diversity." *Nature* 461: 254–7.
- Li, S., J. Tan, X. Yang, C. Ma, and L. Jiang. 2019. "Niche and Fitness Differences Determine Invasion Success and Impact in Laboratory Bacterial Communities." *The ISME Journal* 13: 402–12.
- MacArthur, R. H., and R. Levins. 1967. "The Limiting Similarity, Convergence, and Divergence of Coexisting Species." *American Naturalist* 101: 377–85.
- May, R. M. 1975. *Stability and Complexity in Model Ecosystems*, 2nd ed. Princeton, NJ: Princeton University Press.
- Meine, C. 1999. "It's about Time: Conservation Biology and History." *Conservation Biology* 13(1–3): 2–3.
- Melbourne, B., and A. Hastings. 2008. "Extinction Risk Depends Strongly on Factors Contributing to Stochasticity." *Nature* 454: 100–3.
- Méléard, S., and D. Villemonais. 2012. "Quasi-Stationary Distributions and Population Processes." *Probability Surveys* 9: 340–410.
- Mordecai, E., K. Gross, and C. Mitchell. 2016. "Within-Host Niche Differences and Fitness Trade-Offs Promote Coexistence of Plant Viruses." *The American Naturalist* 187(1): E13–26.
- Muller-Landau, H., and M. Visser. 2019. "How Do Lianas and Vines Influence Competitive Differences and Niche Differences among Tree Species? Concepts and a Case Study in a Tropical Forest." *Journal of Ecology* 107: 1469–81.
- Narwani, A., M. A. Alexandrou, T. H. Oakley, I. T. Carroll, and B. J. Cardinale. 2013. "Experimental Evidence that Evolutionary Relatedness Does Not Affect the Ecological Mechanisms of Coexistence in Freshwater Green Algae." *Ecology Letters* 16: 1373–81.
- Ovaskainen, O., and B. Meerson. 2010. "Stochastic Models of Population Extinction." *Trends in Ecology & Evolution* 25: 643–52.
- Pande, J., T. Fung, R. Chisholm, and N. M. Shnerb. 2020. "Mean Growth Rate when Rare Is Not a Reliable Metric for Persistence of Species." *Ecology Letters* 23: 274–82.
- Pande, J., T. Fung, R. Chisholm, and N. M. Shnerb. 2020. "Invasion Growth Rate and Its Relevance to Persistence: A Response to Technical Comment by Ellner et al." *Ecology Letters* 23: 1725–6.
- Reuter, G. 1961. *Competition Processes* 421–30. Berkeley, CA: University of California Press.
- Saavedra, S., R. P. Rohr, J. Bascompte, O. Godoy, N. J. B. Kraft, and J. M. Levine. 2017. "A Structural Approach for Understanding Multispecies Coexistence." *Ecological Monographs* 87(3): 470–86.
- Schreiber, S. 2022. "R Code for "Does Deterministic Theory Matter in a Finite World?" https://figshare.com/articles/software/R_code_for_Does_deterministic_theory_matter_in_a_finite_world/17125598/1.
- Schreiber, S., M. Benaïm, and K. A. S. Atchadé. 2011. "Persistence in Fluctuating Environments." *Journal of Mathematical Biology* 62: 655–83.
- Schreiber, S. J. 2000. "Criteria for Cr robust permanence." *Journal of Differential Equations* 162: 400–26.
- Schreiber, S. J. 2017. "Coexistence in the Face of Uncertainty." In *Recent Progress and Modern Challenges in Applied Mathematics, Modeling and Computational Science* 349–84. New York, NY: Springer.
- Schreiber, S. J., M. Yamamichi, and S. Y. Strauss. 2019. "When Rarity Has Costs: Coexistence under Positive Frequency-Dependence and Environmental Stochasticity." *Ecology* 100: e02664.
- Shoemaker, L., L. Sullivan, I. Donohue, J. Cabral, R. Williams, M. Mayfield, J. Chase, et al. 2020. "Integrating the Underlying Structure of Stochasticity into Community Ecology." *Ecology* 101: e02922.
- Song, C., G. Barabás, and S. Saavedra. 2019. "On the Consequences of the Interdependence of Stabilizing and Equalizing Mechanisms." *The American Naturalist* 194(5): 627–39.
- Spaak, J., and F. De Laender. 2020. "Intuitive and Broadly Applicable Definitions of Niche and Fitness Differences." *Ecology Letters* 23: 1117–28.
- Stump, S., C. Song, S. Saavedra, J. Levine, and D. Vasseur. 2021. "Synthesizing the Effects of Individual Level Variation on Coexistence." *Ecological Monographs* 92: e1493.
- Vasseur, D., P. Amarasekare, V. Rudolf, and J. Levine. 2011. "Eco-Evolutionary Dynamics Enable Coexistence Via Neighbor-Dependent Selection." *The American Naturalist* 178: E96–E109.

- Vellend, M. 2010. "Conceptual Synthesis in Community Ecology." *Quarterly Review of Biology* 85: 183–206.
- Wainwright, C., J. HilleRisLambers, H. Lai, X. Loy, and M. Mayfield. 2019. "Distinct Responses of Niche and Fitness Differences to Water Availability Underlie Variable Coexistence Outcomes in Semi-Arid Annual Plant Communities." *Journal of Ecology* 107: 293–306.
- Warner, R., and P. Chesson. 1985. "Coexistence Mediated by Recruitment Uctuations: A Field Guide to the Storage Effect." *The American Naturalist* 125(6): 769–87.
- Watkinson, A. 1980. "Density-Dependence in Single-Species Populations of Plants." *Journal of Theoretical Biology* 83(2): 345–57.
- Zepeda, V., and C. Martorell. 2019. "Fluctuation-Independent Niche Differentiation and Relative Non-linearity Drive Coexistence in a Species-Rich Grassland." *Ecology* 100(8): e02726.

SUPPORTING INFORMATION

Additional supporting information can be found online in the Supporting Information section at the end of this article.

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