

MacArthur’s consumer-resource model: a ‘Rosetta Stone’ for competitive interactions

Jawad Sakarchi¹✉ and Rachel Germain¹

¹[Department of Zoology and the Biodiversity Research Centre, the University of British Columbia, Vancouver, British Columbia, Canada]



Recent developments in competition theory, namely, Modern Coexistence Theory (MCT), have aided empiricists in formulating tests of species persistence, coexistence, and evolution from simple to complex community settings. However, the parameters used to predict competitive outcomes, such as interaction coefficients, invasion growth rates, or stabilizing differences, remain biologically opaque, making findings difficult to generalize across ecological settings. Here, our article is structured around five goals, towards clarifying MCT by first making a case for the modern-day utility of MacArthur’s consumer-resource model, a model with surprising complexity and depth: (i) to describe the model in uniquely accessible language, deciphering the mathematics towards cultivating deeper biological intuition about competition’s innerworkings regardless of what empirical toolkit one uses, (ii) to provide translation between biological mechanisms from MacArthur’s model and parameters used to predict coexistence in MCT, (iii) to make explicit important but understated assumptions of MacArthur’s model in plain terms, (iv) provide empirical recommendations, and (v) to examine how key ecological concepts (e.g., r/K selection) can be understood with renewed clarity through MacArthur’s lens. We end by highlighting opportunities to explore mechanisms in tandem with MCT and to compare and translate results across ecological currencies towards a more unified ecological science.

coexistence theory | competition theory | contemporary niche theory | Lotka-Volterra | R^* | resource utilization

Correspondence: sakarchi@zoology.ubc.ca

*****Note: There may be minor discrepancies between this version and the submitted version*****

Introduction

The Rosetta Stone, a slab carved by Egyptian clergy in 196 BC to translate between hieroglyphic, Demotic, and Greek script, was lost for tens of hundreds of years until it was discovered during Napoleon’s invasion of Egypt (Ray 2014). Prior to this discovery, no one knew how to interpret hieroglyphics—hieroglyphics were often interpreted literally, as pictographs representing ideas or events (e.g.,  represents a vulture) rather than symbols representing sounds of language (i.e.,  represents “ah”), and as a result, ancient Egypt remained a mystery. Much like ancient civilizations, ecology is a field of many languages in the form of theoretical frameworks each depicting some aspect of how organisms interact with each other and their environments. Some theories deal with flows of matter and energy (Loreau 1995; Sterner and Elser 2002; Brown et al. 2004), others with population abundances and dynamics (Hutchinson 1978) or the distribution of

species within and among communities (Leibold and Chase 2017). Although these theories are interrelated and sometimes even involve or describe similar variables or phenomena (e.g., low density growth rates (Roff 1993; Arendt 1997; Saavedra et al. 2017)), like languages, each is rich with tradition (e.g., notation style) and history (e.g., how meaning of parameters has changed over time (Barabás et al. 2018)), has a unique system of assumptions and symbolic representations, and is complex in ways that make translation between theories difficult (Grainger et al. 2022; Ou et al. 2022).

To help translation between seemingly disparate theories in ecology, we focus here on deciphering Robert H. MacArthur’s consumer-resource model (equation 1; (MacArthur 1969a, 1970, 1972)) for a general audience while highlighting key misinterpretations. We focus on MacArthur’s model for two reasons. First, we view MacArthur’s model as an intermediary with potential to connect and better understand ecological theories, including modern theories that simplify prediction by skipping the biological details (Box 1). Second, for some, MacArthur’s model is intuitive—it is easy to picture, for example, birds requiring, consuming, and growing populations in response to available seed resources—and thus, can be used to better intuit outcomes in a broad range of ecological scenarios. However, as we will discuss, there are important nuances to interpreting MacArthur’s model that are easy to get wrong in ways that are consequential.

Our article is structured around four primary goals: (i) to describe MacArthur’s model in biological terms using best practices of accessible writing (e.g., numerical examples, relatable analogies (Shoemaker et al. 2021; Grainger et al. 2022; Ou et al. 2022)), (ii) to present a mathematical translation between parameters of MacArthur’s model and parameters more familiar to practitioners of phenomenological models of competition, (iii) to make explicit important (mis)interpretations, assumptions, and limitations, and (iv) provide empirical recommendations. In doing so, we also make a case for the modern-day utility of MacArthur’s model (Box 1) and examine how key ecological concepts (i.e., r/K selection, positive frequency-dependent competition) can be better understood through MacArthur’s lens (Box 2). Importantly, our contribution here is not simply a review of concepts that can be gathered from any handful of source articles or textbooks—it is the result of 5 years of discussion, puzzlement, and discovery. Much of what we say here is a synthesis of material scattered across hundreds of articles published across decades, and throughout, we make explicit

94 interpretations of the model that have only ever been made 143
95 implicitly, shedding new light even on concepts that might 144
96 feel well-understood at first glance. 145

97 (i) MacArthur's consumer-resource model: 147 98 building biological intuition 148

99 To aid our discussions, we necessarily begin by first walk- 150
100 ing the reader through the fundamentals of MacArthur's 151
101 consumer-resource model, first presented in (MacArthur 152
102 1969): 153

$$\frac{1}{X_i} \frac{dX_i}{dt} = C_i \left[\sum_{k=1}^m a_{ik} w_k R_k - T_i \right] \quad (1.1) \quad 154$$

$$\frac{1}{R_k} \frac{dR_k}{dt} = r_k \left[1 - \frac{R_k}{K_k} \right] - \sum_{i=1}^n a_{ik} X_i \quad (1.2) \quad 155$$

103 The model describes dynamics that arise as populations 161
104 of consumer species i of density X_i consume and grow in re- 162
105 sponse to resource species k of density R_k . Two features are 163
106 worth highlighting. First, competition among consumers and 164
107 predator-prey dynamics are modeled simultaneously, unlike 165
108 phenomenological models that must treat these two types of 166
109 interactions separately. Second, this model can be extended 167
110 to any number of consumer species (denoted by n), resource 168
111 species (denoted by m), and in subsequent work by others 169
112 (Chesson and Kuang 2008; McPeck 2022) to more than two 170
113 trophic levels. 171

114 In the absence of consumers (i.e., setting X_i to 0), re- 172
115 sources in equation 1.2 grow logistically; they recover from 173
116 low densities at rates described by r_k , reaching an equilib- 174
117 rium population size at K_k . In the presence of a single in- 175
118 dividual of consumer species i ($X_i = 1$), each individual of 176
119 resource k has a probability of being removed of a_{ik} (the 177
120 per capita attack rate) per unit time. More consumer indi- 178
121 viduals means more drawdown of resources, decreasing the 179
122 resource's per capita growth rate. In equation 1.1, param- 180
123 eter w_k "weights" each resource type by its overall nutri- 181
124 tional quality to consumers, essentially converting resource 182
125 intake into usable energy (or consumer biomass, hencefor- 183
126 ward referred to as 'energy' for brevity). Thus, $\sum a_{ik} w_k R_k$ 184
127 in equation 1.1 represents total energy intake by each indi- 185
128 vidual of consumer species i (on average at an instance of 186
129 time) by consuming resources of different weights, adding to 187
130 growth, whereas term T_i represents each individual's base- 188
131 line energetic requirements (on average at an instance of 189
132 time). Populations grow ($(1/X_i)(dX_i/dt) > 0$) when indi- 190
133 viduals eat more than they require (on average) and shrink 191
134 ($(1/X_i)(dX_i/dt) < 0$) when they require more than they eat. 192
135 The exact rate of population growth or decline is set by C_i , 193
136 the conversion of net energy gain/loss into gain/loss of con- 194
137 sumer individuals. For example, if an individual consumes 195
138 100 more energy units than it needs and each offspring costs 196
139 20 energy units to produce, then the individual can afford to 197
140 produce five offspring. Although this model feels intuitively 198
141 familiar in ways that ease understanding, as we will discuss, 199
142 it harbors a surprising complexity and nuance.

Resource utilization': an ecological hieroglyph One of 200
ecology's most widely used symbolic representations is the 201
Gaussian curve depicting resource utilization, a dimension 202
of the niche that contributes to the outcome of competition. 203
Generally speaking, resource utilization functions depict the 204
ability of an individual, population, or species to consume or 205
usurp certain resource types relative to other resource types 206
(Roughgarden and Feldman 1975). Utilization functions are 207
often drawn cartoonishly to simplify communication of com- 208
plex concepts (e.g., (Araújo et al. 2011; Brodersen et al. 209
2018; Ponisio et al. 2019)) and support verbal arguments 210
in presentations or research papers. For example, in devising 211
predictions for how competition coefficients might change as 212
competing species coevolve in sympatry (i.e., a phenom- 213
ological approach), it can be helpful to consider changes in 214
the strength and direction of selection on resource utilization 215
felt by each species (e.g., figure 1 in (Germain et al. 2020)). 216
However, utilization curves have a precise theoretical mean- 217
ing—we will keep the meaning of "utilization" vague for now 218
because, as we will describe, there are several variants of how 219
utilization has been presented in the literature with important 220
consequences for interpretation. Note that these functions 221
need not be ordered on a one-dimensional axis (i.e., seeds 222
based on size), Gaussian, or continuous: they can take on 223
any shape (although the math becomes more complex (see 224
(Roughgarden 1974; Pigolotti et al. 2010; Leimar et al. 225
2013)), with the area under the curve representing total uti- 226
lization across all resource types (using summations if dis- 227
crete (as per equation 1.1) or integral calculus if continuous). 228
Although equation 1.1 deals with discrete resources, we de- 229
fault to continuous (ordered) Gaussian curves in our figures 230
due to their familiarity to most readers. Here we aim to un- 231
pack different variants of "utilization", their interpretation, 232
how they relate to one another, and streamline inconsisten- 233
cies in terminology. In all cases, the height at each point of 234
each curve in figure 1 depicts the magnitude of a given re- 235
sponse variable for a given resource type.

(a) Attack rate functions Attack rate functions describe how 236
attack rates (a_{ik} in equation 1) are distributed across k re- 237
sources (solid lines in figure 1A,C). These rates are assumed 238
in equation 1 to be constant over time and space (unless spec- 239
ified otherwise) and spatially implicit, meaning that every re- 240
source item in a given unit of space has an equal probability 241
of being consumed by every consumer individual (see table 242
1) — by virtue of describing rates of resource removal, some 243
ecologists prefer to refer to attack rates as '(space) clearance 244
rates' (Narwani and Mazumder 2010; DeLong et al. 2018). 245
Attack rate functions reflect traits (e.g., behavioral, morpho- 246
logical) of both the consumer (e.g., different search strate- 247
gies) and resources (e.g., any countermeasures (Beardsell et 248
al. 2021)). Attack rate functions can be used to predict how a 249
consumer would perform in any hypothetical resource envi- 250
ronment (Lawlor 1980), both alone and in competition with 251
other consumers. For example, a consumer may be capable 252
of consuming a given resource, but if that resource is absent, 253
the consumer will be unable to meet its requirements and will 254
not persist (MacArthur 1969). As such, attack rate functions

Box 1. MacArthur's centrality in modern-day ecology

One might wonder why MacArthur's model is worth revisiting at this point in time. No one would argue against MacArthur's central place in ecology's history—his model is well known (at least in terms of its existence) and is understood in great depth by some subset of ecologists, particularly, some theoreticians (Barabás et al. 2018; Chesson 2020; Abrams 2022). At the same time, many modern-day ecologists consider “classical” mechanistic models like MacArthur's to be antiquated given the empirical intractability of measuring all necessary parameters for all relevant resource dimensions (Letten et al. 2017), to the extent that the current generation of ecologists rarely are exposed to this older body of theory as part of their training. “Modern” Coexistence Theory (MCT) has been offered as a popular solution (Hart et al. 2018; Grainger et al. 2019b). Where competition between consumers in mechanistic models largely emerges as an indirect consequence of resource consumption, MCT instead models competition as though it were a direct interaction (McPeck 2022). In doing so, population-level parameters important to competition (e.g., competition coefficients) can be measured by manipulating densities of consumers (Hart et al. 2018) without needing to know the biological details of what consumers are competing for and how—this is often referred to as a more “phenomenological approach” to studying competition (McPeck 2022). This, however, comes at a cost: we have learned from our own experience that it can be challenging to decipher how parameters in a phenomenological model of competition arise biologically (in general and in a specific experimental context), making it hard to predict how parameters (and thus outcomes of interest) might translate across different ecological settings.

Why now? One might argue that to revisit MacArthur's models is to repeat history (Graham and Dayton 2002). Indeed, MacArthur's motivation for developing a mechanistic theory of competition stemmed from a dissatisfaction with phenomenological models, such as Volterra's equations, which may be “hopelessly far” (MacArthur's words) from the reality of nature in many situations (appendix of chapter 2 in (MacArthur 1972)); this is in part why MacArthur's theory and theories derived from it (e.g., (Tilman 1982)) are referred to as “contemporary niche theory” (Leibold 1995)—at the time, a mechanistic approach was the modern one. How can we make progress if we simply keep repeating ourselves? We argue that timing and packaging is everything (hence, the motivation for this special issue). Now more than ever, empiricists are looking to engage with theory, both to motivate and refine experiments (Hart et al. 2018; Grainger et al. 2019b) and to help solve applied problems (e.g., restoration (Aoyama et al. 2022)). This desire is being met with an interest in making theory accessible to a wider audience (Shoemaker et al. 2021; Ou et al. 2022), especially mathematical representations of theory (Grainger et al. 2022)—this in our opinion has been the greatest roadblock: texts aligning phenomenological and mechanistic approaches to competition theory exist (e.g., (Letten et al. 2017)) but are written at a level that may be difficult for most readers to fully grasp.

At the same time, over the past 50 years, ecology has matured as a more synthetic science; revisiting old models with a new lens and new tools (e.g., computational methods (Bolker 2008), experimental techniques (Jolliffe 2000)) can lead to new research avenues. Importantly, we are not advocating for the re-adoption of mechanistic approaches over phenomenological ones (but see (Abrams 2022; McPeck 2022))—in many ways, the phenomenological approach offered by MCT opens up lines of inquiry that would not be feasible to test otherwise, for example, of how overall competitive ability varies among species based on evolutionary history (Sakarchi and Germain 2023), order of arrival (Grainger et al. 2019a), or environmental context (Lanuza et al. 2018; Van Dyke et al. 2022). Rather, we advocate for a deeper understanding of MacArthur's model so that it could be used to help build biological intuition when using other approaches, such as MCT.

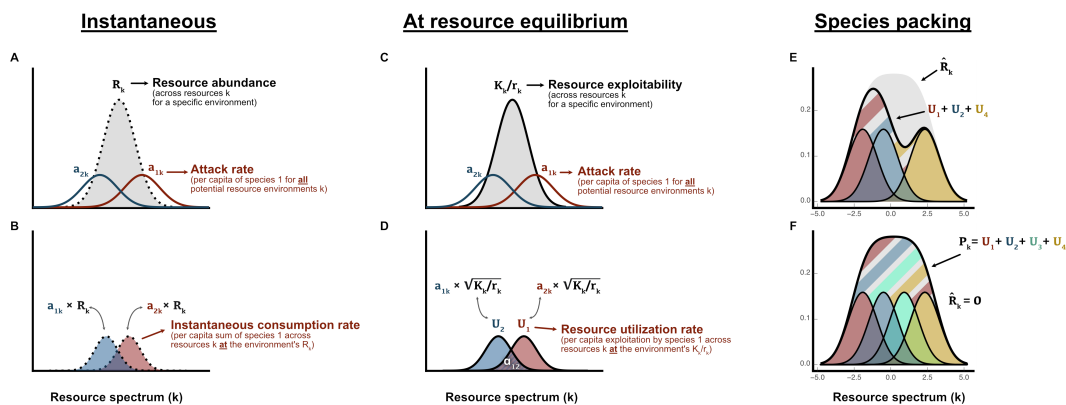


Fig. 1. Clarifying confusion about differences between attack rates, consumption rates, and utilization (panels A-D). Panels (A, C) depict the attack rate functions of two consumers, red (consumer i) and blue (consumer j), for any given level of resource abundance (A), which determines how many resources are consumed by each species at any moment in time (B), or based on relative to rates of resource exploitability (see section i(b)) to determine utilization rates (D). Panel (E) shows an underpacked community where some resources are underutilized, leaving the community susceptible to invasion, whereas panel (F) presents a community of species that fully utilize available resource production (see section i(c)), and thus are fully “packed”. The gap in gray between R_k and the summed utilization functions represents the U component of Q (described section i(c)).

can be used to characterize a consumer’s fundamental and realized niche (Hutchinson 1957; Carscadden et al. 2020), In addition, the degree of overlap in attack rate functions between consumers describes similarities in those consumers’ resource niche, or what Lawlor (1980) refers to as ‘consumer similarity’. Quantifying attack rate functions can help understand phenomena such as niche evolution (MacArthur and Wilson 1967; Lawlor 1980) or potential spread of species to new areas (Usui et al. 2023). Confusingly, attack rate functions are often referred to as ‘utilization functions’ (Levins 1968; Schoener 1974; Chesson 1990; Ackermann and Doebeli 2004). For reasons that will become clear (see subsection (b) ‘utilization functions’ below), attack rate functions are distinct from utilization and cannot be used to predict (i) a consumer’s performance, (ii) how many resources a consumer will consume, or (iii) the outcome of competition in the absence of information about the resource environment. However, as Lawlor (1980) argues, if consumers overlap perfectly in attack rate functions, they will also overlap perfectly in all resource environments given that species will be equivalent regardless of context (whereas the inverse is not necessarily true).

(b) Utilization functions For reasons that will become clear, ‘utilization’ (sometimes referred to as ‘total’, ‘actual’ or ‘weighted’ utilizations (Schoener 1974), represented in MacArthur (1970) and May (1974) by the following equation:

$$U_i = \sum_{k=1}^m a_{ik} \sqrt{w_k K_k / r_k} \quad (2)$$

can be thought of as a consumer’s ability to impact the amount of resources available to other consumers in the immediate future (a description that sounds vague but is theoretically precise)—these rates are not to be confused with the actual amount of resources each consumer eats at a given point in time (calculated as $\sum_{k=1}^m a_{ik} R_k$), a quantity Tilman (1980) instead refers to as the ‘instantaneous consumption rate’ (figure 1B). Although instantaneous consumption rates determine if a population of consumers will instantaneously grow or shrink given resource abundances at a given point in time, they cannot be used to predict whether or not the population will persist or how strongly a population will experience competition. To understand why, consider a consumer entering an environment replete with resources. At first, the population would grow, giving the impression that the population will persist. However, if the resource ceases to be renewed (e.g., a resource patch formed by a pulsed event), resources will eventually be depleted to zero, resulting in the eventual extinction of the consumer.

In contrast to instantaneous consumption rates, utilization is not necessarily based on consumption itself but on the consequences of consumption for future growth in ways that can be used to predict the outcome of competition. Specifically, for utilization, what matters most is how attack rates align with the rate of resource “re-supply” (i.e., the $w_i K_k / r_k$ portion of the utilization equation (figure 1C vs D)). Consider

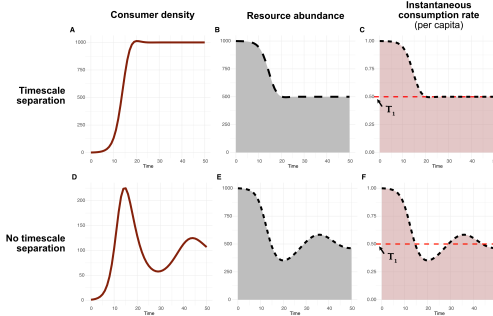


Fig. 2. Comparison of population dynamics of a consumer and its resource with (A-C) and without (D-F) a separation of timescales (see section iii(c)). Panels A-C show the approximate logistic growth of the consumer generated by the timescale separation, modeled by giving the resource a fast low-density growth rate ($r_k = 2$). This fast resource growth corresponds to a consumer utilization (U_{1k}) of 0.5. Panels D-F now show dynamics (e.g., oscillations) when the assumption of timescale separation is broken, with slow resource growth ($r_k = 0.2$) leading to a high consumer utilization ($U_{1k} = 5$). Notably, with timescale separation, resources never fall below a density that fails to meet each individual’s minimum requirements (T_1 ; panels C vs. F). All other parameters are held constant: $K_k = 1000$, $a_{1k} = 0.001$, $T_1 = 0.5$.

two resources: mice and rabbits, which both serve as prey for an aerial predator. Because mice have shorter generation times than rabbits (approximated under continuous time for simplicity), their populations are “re-supplied” at a faster rate ($r_{mice} > r_{rabbit}$), rabbits are unavailable for consumption more frequently. In other words, if prey populations have not recovered from consumption by the time the predator returns, then the predator will experience competition. Thinking about this problem mathematically, imagine a single consumer entering an environment where a resource with a nutritional weight (w_i) of 1 is at its carrying capacity of 800 resource individuals per enclosure. Let’s say one resource is attacked (a_{ik}) with a rate of 0.1 and the rate of resource re-supply (r_k) is 2.0, in other words, the resource is replenished 20x faster than it is removed per unit time. Even though 10% of resource is attacked and used to grow new consumers, after accounting for resource re-supply, only 5% of the 800 resources are unavailable to future consumers, resulting in a utilization of 2. If r_k was instead 1.0 (a slower rate of re-supply), utilization would increase to $\tilde{2}.83$.

The effect of r_k on utilization is important in a competitive context because it means that competition is weaker for resources that are resupplied at faster rates ($r_{mouse} > r_{rabbit}$), i.e., resources that are less scarce or less limiting, which is a central tenet of competition theory (Tilman 1984). More precisely, competition is highest for resources that have what we will call the highest ‘exploitability’ (K_k / r_k). Exploitability corresponds to the time each resource takes to return to its carrying capacity, which is highest for resources that are more abundant (higher K_k) or that are resupplied more slowly (lower r_k). Taken to the extreme, as noted by MacArthur (1972), if resources were to renew instantaneously ($r_k \rightarrow \infty$), consumers would not experience competition. Inversely, if resources were increasingly non-renewing ($r_k \rightarrow 0$, such that $K_k / r_k \rightarrow \infty$), competition would intensify and be destabilized (see section iii(a) about timescale separation) until the resource is entirely expended. Taken

Table 1. Description of parameters in MacArthur's consumer-resource model, including their units, assumptions, and extensions where assumptions are relaxed

Parameter (& Domain)	Example units	Biological definition and comments	Notable assumptions (Special cases are underlined>)	Examples of papers that relax these assumptions
C_i ($0 < C_i < \infty$)	grams _{Daphnia} × Joules ⁻¹	Definition: Conversion efficiency — a constant that turns combined nutritional weight (e.g., Joules or grams) of substitutable resources acquired in excess of requirement to grams of new individuals (offspring). Comments: Often depicted as unitless (as it would be g_i/g_k), unless w_k is better defined with different units like Joules (see w_k)	I. Conversion of energy to grams of individuals is linear (proportional to resources obtained) II. Conversion efficiency is identical between all resources	I. Perrson et al. 1998; Weitz & Levin 2006 II. Leibold 1988; Abrams & Cortez 2015
a_{ik} $0 \leq a_{ik} < \infty$	flask × Daphnia ⁻¹ × Day ⁻¹	Definition: Attack rate represents the probability the consumer i both encounters and consumes resource k per unit time. This can be thought of as what portion of the experimental unit (e.g., "50ml flask") a consumer can encounter and consume a resource in a unit time (day). Comments: If the attack rate is 10% per day, this would represent 0.1 "50ml flask"/day. However, if probability is greater than 100% (i.e. $a_{ik} > 1$), this suggests the consumer would take less than the unit time (one day) to encounter and consume the resource k .	I. Independent of resource density. Representing a Type 1 (linear) functional response (consumption is instantaneous, no handling time) II. Independent of consumer density (no Allee effects, interference, or mutualisms) III. Independent of environmental conditions (temperature independent) IV. Resource competition occurs only over (constantly) limiting resources V. Attack rate is constant over time (no optimal foraging) VI. Resources depleted by consumers are entirely used for consumer growth (no wastage)	I. Stewart & Levin 1973; Real 1977; Armstrong & McGehee 1978, Abrams 1980 II. Beddington 1975; DeAngelis et al. 1975; Holland & DeAngelis 2010 III. Betini et al. 2019; Vasseur 2020 IV. Mazancourt & Schwartz 2012 V. Wiens 1977; Persson et al. 1998; Orlando et al. 2012 VI. (Hernández-García et al. 2009; Leimar et al. 2013)
w_k $-\infty < w_k < \infty$	Joules × Plankton ⁻¹	Definition: Nutritive weight of a specific resource, often considered to be its 'quality' or 'nourishing value' (not to be conflated with C_i , which is a constant conversion rate across all resources). Comments: Usually represented in physical weight (g_k/N_k) but can be converted to other units (e.g., Joules, mg of a specific micronutrient) if physical weights of different resources are not equally nutritious. This would carry over to the conversion efficiency to become a unit-less constant (g_i/g_k). When negative, $w_k < 0$, the resource requires more energy to consume than it provides.	I. Each resource is equal in 'nutritive weight' to all consumers	I. Schoener 1974
R_k $0 < R_k < \infty$	Plankton × flask ⁻¹	Definition: Abundance (density) of resource k , usually in units of either individuals or biomass per unit area. Comments: Sometimes the units are presented without an explicit spatial extent, in which case the spatial unit is equivalent to the experimental unit (e.g., enclosure, pond, island, erlenmeyer flask)—or with explicit space (square meters, gallons). This also extends to the concept of carrying capacity — capturing the 'density' of a population per unit area.	I. Resources are depletable II. Resources are equally accessible (resources and consumers are homogeneously distributed in space) III. Resources are not interacting IV. <u>Resources cannot go extinct if used for Q or L-V translation</u> V. Resources renew logistically VI. Resources are defined per unit of space VII. <u>Resource and consumer dynamics operate on separate time scales if used for Q or L-V translation</u> VIII. <u>If Gaussian utilization functions are used, resources can be ordered along one dimension</u>	II. Sasaki 1996 III. MacArthur 1970; Levine 1976 IV. Hsu & Hubbell 1979; Abrams 1980 V. MacArthur 1972; Abrams 1980 VII. O'Dwyer 2018
T_i $0 < T_i < \infty$	Joules × Daphnia ⁻¹ × Day ⁻¹	Definition: Per capita resource requirement for consumer i ; if resource intake just meets this requirement, the population will not grow or shrink. Comments: Can be thought of as the metabolic requirement to maintain one individual per unit time. This includes the physiological resource requirements for that individual and also demographic turnover, i.e., resources needed to replace death with birth per unit day. "T" stands for threshold.	I. Resource requirement is density independent	I. Schoener 1974; Holt 1985; Aubier 2020

Note: Citations in this table are relegated to the Supplementary Materials due to space constraints. Although different symbols for MacArthur's consumer-resource model are used by different authors (e.g., m_i "mortality" instead of T_i and c_{ij} instead of a_{ij} in Chesson (1990)), we adhere to MacArthur's original symbology given that inconsistency across the literature can impede understanding. Scripts, i , j , and k refer to consumer species, heterospecific consumers in the Lotka-Volterra equations, and resource species, respectively, each with a total species pool of n and m .

290 together, most ecologists recognize that resource availabil- 340
 291 ity (perhaps in a vague sense) is key to determining the out- 341
 292 come of competition. Here, we make the notion of ‘availabil- 342
 293 ity’ more precise in terms of resource exploitability, which 343
 294 emphasizes the importance of dynamic fluxes of resources 344
 295 as opposed to standing stocks. In doing so, we better clar- 345
 296 ify what competition is fundamentally: it is not simply the 346
 297 impact consumers have on each other’s growth, but rather, 347
 298 how rates of consumption affect the amount of time a renew-
 299 able resource is made unavailable to future consumers. Note
 300 that some authors (Ackermann and Doebeli 2004; Abrams et
 301 al. 2008) use K_k instead of K_k/r_k to calculate utilization, 349
 302 which does not aim to diminish the importance of re-supply, 350
 303 but rather, makes the simplifying assumption that resources 351
 304 have equal r_k . 352

305 **(c) Community utilization** MacArthur’s interest in character- 354
 306 izing resource utilization also served an additional purpose: 355
 307 to formulate his ‘minimization principle’ (MacArthur 1969; 356
 308 MacArthur 1970). MacArthur’s minimization principle 357
 309 states that communities should assemble such that resource 358
 310 competition (and its evolution) minimizes inefficiencies in 359
 311 resource utilization at the community level. He labeled these 360
 312 inefficiencies Q , a quantity that MacArthur himself admitted 361
 313 is “not easy to interpret biologically” (MacArthur 1970): 362

$$Q = \underbrace{\sum_k \frac{w_k K_k}{r_k} \left[r_k - \sum_j a_{jk} X_j \right]}_{\mathcal{U}}^2 + 2 \underbrace{\sum_j T_j X_j}_{\mathcal{B}} \quad (3)$$

314 This equation has been summarized by others (Gatto 370
 315 1990; Ghedini et al. 2018)) as a sum of two parts: the unuti- 371
 316 lized productivity of the environment (\mathcal{U} , not to be confused 372
 317 with utilization ‘ U ’ in (2)) and productivity lost to the basal 373
 318 metabolic maintenance of the consumer population (\mathcal{B}). Bro- 374
 319 ken down in into \mathcal{U} and \mathcal{B} , it becomes clearer that Q repre- 375
 320 sents inefficiencies in energy flow from lower trophic levels 376
 321 to higher ones: if a population of consumers used resources 377
 322 in their entirety, consuming new resources as fast as they 378
 323 are produced (i.e., when $a_{jk} X_j = r_k$, such that $R_k \rightarrow 0$ and 379
 324 $\mathcal{U} = 0$; figure 1F), and had infinitesimally small metabolic re- 380
 325 quirements (i.e., $T_i \rightarrow 0$, causing $\mathcal{B} = 0$), then inefficiencies 381
 326 (i.e., Q) would be 0. Of course, it is biologically impossible 382
 327 for populations to subsist with next to zero requirements, and 383
 328 more efficiently utilizing productivity (decreasing \mathcal{U}) would 384
 329 typically come at the cost of increasing metabolic require- 385
 330 ments (increasing \mathcal{B})—thus, every community is inefficient 386
 331 to some degree. 387

332 Q is ecologically important for several reasons. First, if 388
 333 Q decreases over time ($dQ/dt < 0$) it tells us that the com- 389
 334 munity of consumers is not at an equilibrium ($X_j \neq X_j^*$).
 335 Second, and perhaps more importantly, if the community is at
 336 an equilibrium and Q is still not minimized (i.e., alternative
 337 combinations of species that reduce Q further are biologi-
 338 cally possible; gray portion of figure 1E), then the community 390
 339 is susceptible to invasion by new species or new genotypes — 391

by “biologically possible”, we mean species that are closer to
 an optimized form given realistic biological constraints — for
 example, a species with zero metabolic requirements ($T_i = 0$)
 is biologically impossible. In other words, Q is closely tied
 to conditions for coexistence in multi-species communities
 (MacArthur 1969). Additional details on species packing
 with an emphasis on communities assembled by evolution
 can be found in Germain et al. (2024).

(ii) Translating MacArthur to Lotka-Volterra

Modern coexistence theory has become a popular tool for
 testing hypotheses of how species persist, coexist, and evolve
 in simple to complex community settings, however, there re-
 mains concern over how phenomenological ‘mechanisms’ of
 coexistence underlying modern coexistence theory can be in-
 terpreted biologically (Box 1 (Abrams 2022; McPeck 2022)).
 We argue that these concerns could be addressed in part by
 making explicit the mechanistic underpinnings of the param-
 eters found in phenomenological models of coexistence, both
 verbally and mathematically (table 2), which we provide here
 in one place for the first time. Although these mathematical
 translations can be found scattered throughout the classical
 literature, they are easy to miss—for example, MacArthur
 (1970) represents a consumer’s intrinsic rate of increase us-
 ing \mathcal{K} , and nowhere in that paper does he actually refer to \mathcal{K}
 as the intrinsic rates of increase. In addition, MacArthur was
 a champion of brevity, presenting the equations with little bi-
 ological interpretation. We focus on the Lotka-Volterra com-
 petition model given its widespread use in ecology, including
 in the coexistence literature (e.g., (Broekman et al. 2019)) —
 other commonly encountered models, such as the Beverton-
 Holt annual plant model, make slightly different assumptions
 but share many general features with Lotka-Volterra (Weigelt
 and Jolliffe 2003). By providing a translation to mechanisms,
 this section will make clear two important take-home mes-
 sages about the Lotka-Volterra competition model: (1) that
 its parameters (e.g., r , α , and K) are not independent, as they
 share mechanistic underpinnings, and (2) that every param-
 eter is affected by the resource environment (albeit in different
 ways). To avoid confusion, symbols ‘ r ’ and ‘ K ’ will con-
 tinue to represent the **resources’** intrinsic growth rate and
 carrying capacity (as per equation (1.1)); to clearly distin-
 guish the consumer from the resource, we will use symbols
 ‘ s ’ and ‘ L ’ to refer to the intrinsic growth rate and carrying
 capacity of the **consumer**, respectively (note: ‘ s ’ and ‘ L ’ are
 one-letter. shift from ‘ r ’ and ‘ K ’).

We begin by walking through the Lotka-Volterra compe-
 tition model (see table 2 for a quick reference of parameters),
 a model which describes the growth of different consumer
 species i as they compete with conspecifics (i.e., X_i) and het-
 erospecifics (i.e., X_j):

$$\frac{1}{X_i} \frac{dX_i}{dt} = s_i \left[1 - \frac{X_i}{L_i} - \frac{\sum_{j=1}^n a_{ij} X_j}{L_i} \right] \quad (4)$$

This version of the Lotka-Volterra equation forces the in-
 traspecific interaction coefficient (α_{ii}) to be 1, thus, α_{ij} is

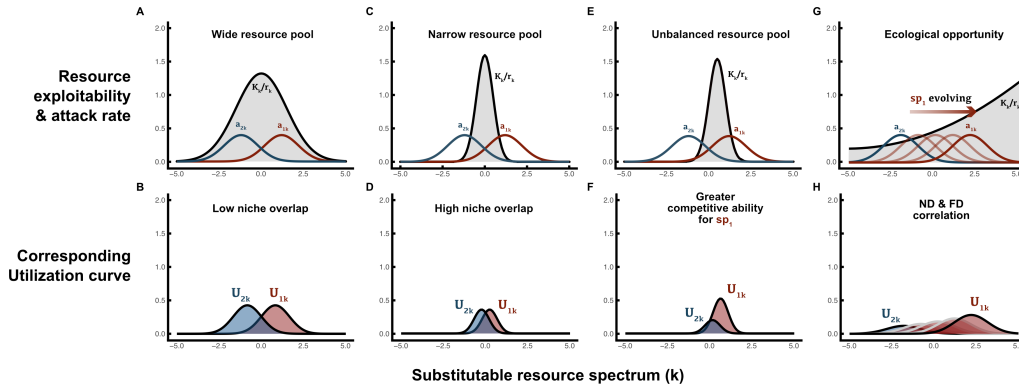


Fig. 3. Effect of the resource environment on competitive differences among consumers. Panels (A,C,E) depict the same two consumers, red (consumer i) and blue (j), in three different years. In years with a wide and more even distribution of resources (A), utilization closely mirrors the consumer’s attack rate function (B). By contrast, differences among consumers collapse (i.e., increased overlap (D)) or result in a competitive asymmetry (i.e., an asymmetry in total utilization (F)) when resources are more narrowly distributed (C) and favor one consumer (E). Panels (G,H) illustrate how the resource environment can create ecological opportunities that consumers can evolve to better exploit (i.e., a shift in the red consumer’s attack rate function), here, illustrating how competitive asymmetries (“fitness differences”, or FD, to mirror the MCT literature) and niche partitioning (“niche differences”, or ND) can evolve to increase in tandem.

relative to α_{ii} . As a consequence, a necessary condition for coexistence in this model is that intraspecific competition must be greater than interspecific competition; this condition is met when all $\alpha_{ij} < 1$.

First we examine Lotka-Volterra parameter s_i (equation (4), table 2(A)), which describes how quickly a consumer population can grow from low densities (i.e., when all $X \approx 0$). Intuitively, s_i should increase with resource densities, but why and by how much? When $X \approx 0$, resource densities grow to carrying capacity (K_k) as no consumption takes place. As such, s_i can be calculated by setting R_k in equation 1.1 to K_k . Doing so means that energy available for growth is simply the difference between energy gained by consumption ($\sum a_{ik} w_k K_k$) and energy lost to metabolism (T_i). Parameter C_i converts this excess energy gain or loss into gain or loss of consumer individuals. So, the consumer’s growth rate at low densities (s_i) is higher in environments where resources have a higher K_k or nutritional value (w_k) or for consumers with higher overall attack rates (a_{ik}), lower metabolic requirements (T_i), or a higher conversion efficiency (C_i); s_i is unaffected by rates of resource resupply.

As consumer populations grow, the Lotka-Volterra competition model includes two quantities that describe how consumers impact each other’s growth as consumer densities increase: L_i and α_{ij} . Consumers reach carrying capacity (L_i) when resources are drawn down to a density low enough that any energy captured ($\sum a_{ik} w_k R_k$) is just enough to meet metabolic requirements (T_i). When assessed separately for each resource species k , the exact resource density corresponding to L_i is the consumer’s R^* for that resource, calculated as $T_i / (a_{ik} w_k)$ (see table 2E for calculation of R^* when multiple resource species are present). R^* is also the resource’s equilibrium density when the consumer is at L_i — i.e., it represents the standing productivity of the system that cannot be used by consumers, which helps illustrate exactly why consumers with higher requirements (T_i) are inefficient from an energetic perspective (see section i(c)). At this R^* , because the resource is below its own carrying capacity

(i.e., $R_k < K_k$), the resource begins every moment in time by growing logistically, producing what MacArthur referred to as “useful production”. (table 2G (MacArthur 1972)). This productivity is “useful” in the sense that 100% of resource growth will be consumed to maintain the consumer at its current density. Note that “useful production” is not to be confused with resource exploitability (equation (2)), as the latter is an intrinsic characteristic of the resource agnostic to consumer densities. By substituting R^* into the left-hand side of the equation shown in table 2F, we can solve for X_i , which is the consumer density that corresponds to resources at R^* (table 2B)—this solution is the equation for the consumer’s carrying capacity from Lotka-Volterra except represented by the parameters of MacArthur’s model.

α_{ij} describes the interspecific interaction coefficient, or, the change in a consumer i ’s per capita population growth rate for each individual of consumer j added—resource competition results in positive values of α_{ij} . In MacArthur’s model, interspecific competition is described by the equation shown in table 2C, or how much of an individual’s potential utilization will be usurped by an individual of consumer j (i.e., another species, numerator) relative to an individual of consumer i (i.e., its own species, denominator). As a side note, unlike MacArthur, many ecologists prefer to work with the Verhulst version of the Lotka–Volterra model (Mallet 2012; Broekman et al. 2019):

$$\frac{1}{X_i} \frac{dX_i}{dt} = s_i - \alpha_{ii} X_i - \sum_{j=1}^n \alpha_{ij} X_j \quad (5)$$

In this version, L_i emerges naturally as intraspecific competition (α_{ii}) causes per capita growth rates to decrease from s_i to 0. In this version, α_{ii} can take on any value and α_{ij} is no longer relativized by α_{ii} .

Together, s_i , L_i , α_{ii} , and α_{ij} provide everything one needs to calculate Modern Coexistence Theory’s mechanisms of coexistence, for example in non-fluctuating environments, niche differences that act to stabilize coexistence and competitive ability differences that destabilize it. From our descrip-

tions above, we can see why these two mechanisms may not necessarily vary independently of one another, as has been demonstrated by others (Kuno 1991; Mallet 2012; Barabás et al. 2018; Song et al. 2019). For example, imagine a scenario where a new beneficial mutation rises to fixation in a population that allows for more efficient resource extraction across all resource types. This singular change would lead to an increase in s_i as each individual would have access to more resources at low densities at the cost of simultaneously increasing the intensity of intraspecific interactions, causing shifts in the relative strengths of α_{ij} and α_{ji} —this specific scenario would mean that competition is strongest in ideal environments, strengthening competition-environment covariance, a key contributor to coexistence in fluctuating environments (Johnson and Hastings 2022). Note, however, that even though niche and competitive ability differences are not mechanistically independent, they can still vary in an uncorrelated fashion, as has been obvious in empirical research (Kraft et al. 2015; Buche et al. 2022; Yan et al. 2022) given the diversity of ways mechanistic parameters may change across species and environments (Vincent et al. 1996; Abrams 2022). Even more, although changes in the resource environment can simultaneously impact interaction coefficients and s_i (e.g., if K_k increases), this is not always the case—the rate of resource re-supply has no impact on s_i whereas it is key to the interaction coefficients (so long as K_k is unchanging).

One of the biggest limitations of Modern Coexistence Theory is how sensitive empirical outcomes are to environmental context. Consider how competitive differences between a pair of species change among years depending on interannual variation in resource exploitability (figure 2), even though the species have fixed attack rate functions. In a year with a broad pool of resources, each species’ utilization (figure 2B) largely mirrors its attack rate function (figure 2A)—each species utilizes a similar total amount of resources (i.e., areas under the curves are equivalent), meaning that their competitive abilities are similar (MacArthur and Levins 1967), and the area of overlap is small, suggesting the presence of niche differences. If this pool of resources narrows (figure 2C) or becomes biased in favor of one species (i.e., species j ; figure 2E), coexistence becomes less stable as niche differences collapse (figure 2D) and competitive asymmetries emerge (figure 2F). Often, empirical tests of Modern Coexistence Theory lack information about the resource environment or about how species are using or requiring resources, leading to a common conclusion that responses to, say, an environmental manipulation, are “idiosyncratic” across species (Germain et al. 2018). In other words, Modern Coexistence Theory can reveal how species competitive hierarchies shift in different contexts (experimental or natural), but without knowing why they shift, predicting and forecasting outcomes in new contexts, such as when environments change or when new species are added, is not possible.

(iii) Less understood but important assumptions of MacArthur’s model

As with any model, MacArthur’s model is replete with assumptions. Exploring the consequences of breaking these assumptions has been an enduring focus of empirical (Ayala et al. 1973; Miller et al. 2005) and theoretical (e.g., (Fronhofer et al. 2018)) research, including subsequent work by MacArthur (MacArthur 1972) and Tilman’s well-known exposition in the 1980s (Tilman 1980). Assumptions are important to be aware of because they tend to constrain a model’s generality, for example, of what kinds of study systems the model can and cannot be applied to. Assumptions may apply to individual parameters in the model (see table 1), to the model as a whole, and to the techniques used to analyze the model to arrive at analytical solutions. However, assumptions are not always easy to intuit unless they are explicitly and unambiguously stated (Ou et al. 2022), although an intuition does become easier as one becomes more familiar with a model, an aim of this article. Here, we do not wish to provide an exhaustive review of the model’s assumptions given the availability of other excellent reviews (but see table 1, where we have compiled the most extensive list of assumptions of MacArthur’s model to date). Rather, we wish to highlight three consequential but lesser known or understood assumptions.

(a) A complete separation of timescales (also referred to as a ‘quasi-steady state’ assumption)

An important assumption MacArthur uses to analyze his model is a complete separation of timescales between consumer and resource dynamics (MacArthur 1970). Although many articles state this assumption, descriptions of what it means, why it is needed, and what its consequences are remain difficult to interpret biologically. Timescale separation means that, at every point in time, resources are assumed to start at a density that is in equilibrium with the density of consumers. At this resource density, rates of resource growth (the first term in equation 1.2) are exactly equal to rates of resource removal by consumers (the second term in equation 1.2). We present the equation to solve for this density, which we call \hat{R}_k , in table 2F. Importantly, consumer densities are still free to change, thus, \hat{R}_k exists in a quasi-steady state, constantly adjusting to track the consumer population (figure 3B). If consumers were to instantaneously vanish from a system, the resource population would reach its carrying capacity effectively instantly regardless of the resource’s density when consumers were present.

Assuming a complete separation of timescales is made purely for mathematical ease, although theoreticians argue that it is biologically reasonable in systems where resources have short generations relative to consumers (e.g., plankton growing faster than planktivores). Without this separation, consumer-resource interactions become more complex (Reynolds and Brassil 2013; O’Dwyer 2018), generating consumer dynamics that no longer follow logistic growth (e.g., predator-prey cycling as seen with lynx and hare; figure 3A,B vs 3D,E) — biologically this is not a prob-

lem, but mathematically it complicates attempts to analyze MacArthur’s model, for example, when attempting to align parameters from MacArthur’s model with parameters from the Lotka-Volterra competition model as shown in section (ii) (O’Dwyer 2018). Additionally, without timescale separation, resources may be consumed to extinction, whereas this is much less likely with timescale separation (i.e., resource abundances bottom out as consumers reach their equilibrium).

(b) Resources are substitutable Although rarely explicitly stated, MacArthur’s model only applies to substitutable resources. Non-substitutable (or ‘essential’) resources refer to different resource types that cannot be consumed interchangeably to meet a similar metabolic need. For example, a plant requires both water and nitrogen—consuming more nitrogen does not make up for any shortcomings in the supply of water. By contrast, substitutable (‘non-essential’) resources can be consumed interchangeably. Resources need not have the same nutritional quality in order to be considered substitutable, for example, if a hawk can simply eat more mice to make up for a lack of rabbits. Note that equation 1.1 only has a single requirement term, T_i , for all resources; each resource type contributes to meeting this T_i , which would not be the case for non-substitutable resources.

To extend MacArthur’s model to include non-substitutable resources, we envision that each unique class of non-substitutable resources (e.g., food, water, shelter) represents a non-substitutable resource dimension (d) that may still be composed of substitutable subtypes. Equation (1.1) then becomes:

$$\frac{1}{X_i} \frac{dX_i}{dt} = C_i \min_d \left(\sum_{k=1}^m a_{ikd} w_{kd} R_{kd} - T_{id} \right) \quad (6)$$

In this way, the equation applies Liebig’s law of the minimum (von Liebig 1855), meaning that growth is dictated by the resource dimension that is least available ($\sum a_{ikd} w_{kd} R_{kd}$) relative to the consumer’s requirement (T_{id} ; hence the \min_d). In practical terms, equation (6) allows us to holistically consider a wider array of resources and predict if changes in abundance of any one resource within a given dimension will ultimately influence competition.

(c) No resources are wasted MacArthur’s model implicitly assumes that resource consumption always contributes to consumer growth—in nature, however, resources are sometimes wasted (i.e., made unavailable by a consumer without contributing to its growth). Resource wastage can occur via many mechanisms. For example, brown bears have been known to only consume the fattiest parts of a salmon, leaving 75% of the body that is now unusable to other consumers of live salmon (Lincoln and Quinn 2018). Alternatively, resource wastage may also take the form of interference competition, including territoriality, allelopathy, environmental modification via metabolic, or apparent competition for ‘enemy free space’ (MacArthur and Levins 1967; Holt 1977; Jeffries and Lawton 1984). Here, a consumer may monopolize

resources in a territory even if the consumer is that it is unable to consume those resources (Leimar et al. 2013).

Resource wastage has interesting consequences for competition. For example, wastage might cause a consumer (say, consumer i) to increase its impact on another (consumer j), increasing α_{ij} without corresponding increases α_{ji} or α_{ii} . In other words, in this example, consumer i ’s response to (α_{ji}) and effect on (α_{ij}) consumer j become decoupled (Goldberg 1990; Leibold 1995; Hart et al. 2018; Sakarchi and Germain 2023) while tipping the balance between the relative magnitudes of intraspecific and interspecific competition, creating a competitive asymmetry (but see figure 4 for an alternative scenario). For this reason, several authors (Hernández-García et al. 2009; Leimar et al. 2013) have proposed that, when necessary, utilization functions can be decomposed into two parts (figure 4): ‘effective utilization’, which represents the fraction of utilization that contributes to growth, and ‘total utilization’, which also includes wastage.

(iv) Empirical recommendations

Throughout this article, we have shown how considering resource dynamics in tandem with consumer dynamics can reframe our biological intuition about the inner workings of competition. Here, we highlight fruitful applications and avenues for empirical research inspired by these intuitions.

Strengthening hypotheses: As ecological theory becomes increasingly abstract, we risk detaching our decisions of what we manipulate and measure in experiments from the underlying biology. As one example, competitors might coevolve to use different resources (character divergence) or to use the same resources (character convergence), or might not evolve at all. Although all are theoretically possible, we cannot predict which outcomes will be most likely without considering the dynamics of the resource environment, as these outcomes critically depend on how rates of resupply vary among resource types to shape ecological opportunities (Germain et al. 2024). As another example, one might want to test whether a focal species can persist in different parts of a landscape. However, if resources in that landscape were deposited in a pulsed, periodic fashion, or if the resources with a very slow resupply rate have had a long time to accumulate before the experiment was initiated, the species may at first appear to persist only until resources are wholly depleted, spurring a crash (as discussed in section (i)). Such scenario illustrates what we will call a “pseudo-source” population, the inverse of a “pseudo-sink” (Holt 2009). Similarly, we might draw stronger inferences about the likely winners and losers of competition based on experimental manipulations of, say, increased resource supply (as in figure 2) or temperature (that increases metabolic demands); this contrasts studies that instead conclude that competitive responses are “idiosyncratic” across species, as we describe in section (ii). Lastly, predicting the likely consequences of broken assumptions (listed in table 1) is itself the basis for entire research programs, for example, to test how often resources are wasted, in what ways, and with what consequences for competitive outcomes (sec-

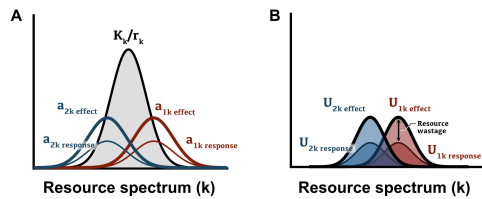


Fig. 4. How resource wastage affects competition. Here, we now show two kinds of attack rate and utilization functions. The curves with thinner lines represent what we refer to here as “effective” attack rate/utilization functions, i.e., resources that are consumed and actually contribute to a consumer’s growth—these are no different from the regular curves presented in figure 1A. By contrast, the curves with the thicker lines represent “total” attack rates/utilization, which includes the “effective component” as well as extra resources impacted through wastage (see section iii(c))—thus, wastage is the difference between these two curves. If only one species wastes resources, this can cause an asymmetry in competitive impact even if those consumers are otherwise competitively equivalent. This figure and the effective vs. total terminology is summarized from Leimar et al. (2013).

tion (iii)). In Box 2, as examples, we unpack two specific research topics in ecology that can be understood with renewed clarity through the lens of MacArthur’s model. This is where we feel this paper will make its biggest impact: by cultivating a stronger biological intuition for the mechanisms underlying competition, stronger hypotheses will naturally follow.

Converting biological currencies across scales: Because MacArthur’s model describes the consumption and conversion of exploitable resources into more consumers, it provides a natural link between biological scales: lower levels, such as metabolism/energetics and stoichiometry of individuals (which includes movement and behavior), to higher levels, such as communities or ecosystems. Reflecting our own expertise, here we have focused on the upwards translation: from individuals to populations to communities. However, we believe that the deeper understanding of MacArthur’s model that our article provides should similarly ease future translations by others to lower levels (e.g., the metabolic, behavioral, and biophysical traits that determine the parameters in MacArthur’s model). Once those connections are made, we imagine being able to examine the cascading consequences of changes at any level to all others, analytically, or more practically, using individual-based simulations. Doing so would reveal the biological cause of fundamental constraints and tradeoffs in biological responses.

Modifying competition experiments to estimate consumer traits and resource dynamics: A common approach to parameterizing phenomenological models of competition is to measure the growth rates of competitors when introduced at low densities in an environment that either lacks competitors (to isolate s_i in equation 4) or that contains resident species at an equilibrium—this approach is called an “invasibility trial” (Grainger et al. 2022). For organisms with simple resource requirements, such as flour beetles, these experiments present an unappreciated opportunity: to estimate each consumer’s attack rates (a_{ik} in equation 1) and resource requirements (T_i) by explicitly considering resource dynamics in tandem with population dynamics. Specifically, when consumers are at their carrying capacity, which is usually the case for the resident population in an invasibility trial, the rate of the to-

tal resource supply is equal to $X_i T_i$ (i.e., the total amount of resources required to sustain the population of consumers at their equilibrium size). In other words, by dividing the rate of total resource resupply by the size of the consumer population, we can estimate T_i . If the experimenter is also able to estimate R_k at this equilibrium, then a_{ik} can be approximated by dividing T_i by R_k (by rearranging equation E in table 2 assuming a linear functional response). In the interest of brevity, in the Supplementary Materials, we provide two detailed case studies of how an experiment could be designed to estimate T_i and a_{ik} . There are many reasons why an ecologist might be interested in a consumer’s resource requirements and attack rates. For example, the distribution of attack rates across resource types constitutes the fundamental niche (section i(a)), and trade-offs between resource requirement and attack rates may represent a fundamental constraint on resource acquisition and coexistence (Box 2(a)).

Conclusions

Here we have showcased Robert MacArthur’s consumer-resource model’s centrality to the theory of competitive interactions and its connection to phenomenological models commonly used to quantify competition and predict coexistence.

In doing so, we:

- made clear the distinction between attack rates, consumption rates, and utilization, quantities that have been conflated in the literature for >40 years
- moved beyond vague notions of “resource availability”, emphasizing “resource exploitability” as a key determinant of competitive outcomes
- clarified exactly what competition is (i.e., the action of making a renewable resource unavailable for a prolonged period of time)
- showed when and why coexistence outcomes are expected to change across resource environments, and sometimes might even give the appearance that species are ecologically equivalent in certain resource environments despite underlying differences in resource niches among species (figure 2)
- made accessible complex concepts, such as timescale separation, “useful production”, and MacArthur’s minimization principle (and the meaning of Q (eq. 4), that are of broad importance to biodiversity studies

As described in Box 1, two excellent books have highlighted a range of concerns (Abrams 2022; McPeck 2022) which they use to stake their opposition to phenomenological approaches. Our stance is more balanced (Germain et al. 2022), instead providing a guide for researchers to endeavor to understand mechanistic models enough to guide hypotheses, to think through assumptions, and to potentially stimulate new research at the intersection of approaches and biological scales.

ACKNOWLEDGEMENTS

We thank current and past members of the Germain lab for years of discussions and Amy Angert, Jennifer Williams, Chris Klausmeier, Robin Snyder, Volker Rudolf, two anonymous reviewers, and the BRC’s “LETS” discussion group for comments on earlier versions of this manuscript.

Box 2. Resolving misconceptions about core ecological concepts through the lens of MacArthur's model

Ecology is a science of how organisms and their environments interact, however, linkages between many concepts and mechanisms that underlie them is not always obvious. Here, we expand upon two core concepts, describing how common misconceptions of each can be resolved by viewing them through the lens of MacArthur's consumer-resource model. Of course, there are many concepts in ecology—we view these specific (and closely related) two as examples of how the inner workings and applications of core concepts could be more easily intuited by being placed within MacArthur's model.

(a) *r/K life history theory* “r/K selection”, as originally envisioned by MacArthur and Wilson (MacArthur and Wilson 1967), refers to the selective environment populations experience when they are kept at densities far below carrying capacity (i.e., in their exponential “r” phase of growth) vs in competition at carrying capacity (i.e., the “K” phase of growth). Instead of focusing on traits, such as investment in many small vs. few large offspring (Pianka 1970); (Reznick et al. 2002), it is useful and perhaps more general to consider how these life history syndromes map onto resource utilization (MacArthur and Wilson 1967). Importantly, increasing overall utilization of resources comes at a cost, as individuals are burdened with increased resource requirements (i.e., higher T_i) to fuel the machinery and increased energetic demands of more efficiently capturing resources (e.g., through enhanced mobility). Consider a population at low density, nowhere near its carrying capacity. Each individual has access to a surplus of resources, even with very low effort. Although increasing utilization efficiency would further increase access each individual has to resources, net fitness benefits may be balanced out by the costs of increasing requirement, and thus, may not evolve. At carrying capacity, individuals are still bound by these tradeoffs. However, even if an increase in utilization comes at the cost of increased requirement, there remains a net benefit as these individuals would outcompete others for scarce resources. By rooting r/K selection by MacArthur's model, several misconceptions become more apparent. First, r-selection only makes sense in environments where populations fall far below K_i , for example, due to disturbances or predators (Reznick et al. 2002) and not environments that simply cannot support a high population density (Taylor et al. 1990). Second, numerous ecology textbooks define K-selection as selection for increasing population size over time as populations adapt to competition (Vandermeer and Goldberg 2013; Hutchings 2021), suggesting that K-selected species interact less strongly with each other compared to r-selected species (that simply never reach high enough densities to compete), a statement also found in the primary literature (Roughgarden 1971)—this seems backwards. Instead, one might expect carrying capacity to shrink over time as populations adapt to increase their attack rate (a_{ik} , thereby lowering K_i , table 2B given that individuals that capture more resources will outcompete others; this creates a Red Queen dynamic favoring the evolution of increased competitive ability, or more generally, a “tragedy of the commons” over shared resources. A systematic review of the r/K-selection literature with a focus on studies that directly measure the evolution of carrying capacities (as opposed to growth rates or selection gradients, as is most common) are needed to weigh evidence for or against these expectations. We believe that these misconceptions have arisen in part due to drift in terminology: that the term ‘K-selected’ has been interpreted as meaning selection **for** K_i under competition, as opposed to referring to populations evolving **at** K_i , as MacArthur initially envisioned when he first coined these terms (MacArthur 1962) and expanded upon in MacArthur & Wilson (1967) in terms of resource utilization. These misconceptions might also explain why r/K selection has fallen out of fashion.

(b) *Positive frequency-dependent competition* An interesting competitive dynamic arises when heterospecifics compete more strongly than conspecifics (Ke and Letten 2018): species increasingly gain a fitness advantage as they outnumber their competitors, resulting in positive frequency-dependent population growth that leads to exclusion of the rarer species. As a consequence, competition may result in alternative stable states depending on which species is able to gain a numerical advantage, for example, by arriving first or due to random fluctuations in density, a dynamic sometimes referred to a ‘priority effect’ in the coexistence literature (but see (Zou and Rudolf 2023) for a commentary on terminology). How can positive frequency-dependence arise in MacArthur's mode? Tilman proposed that consumers may sometimes impact resources they do not strongly require (D. Tilman 1982). Positive frequency dependence then arises when competing species reciprocally impact resources their competitors require most, relative to resources they themselves require. Why would an individual invest time and energy impacting resources for which it does not strongly require? One quickly finds that this kind of dynamic cannot be represented using simple niche cartoons, in contrast to coexistence or competitive asymmetries (e.g., as per figure 2). We contend that ‘resource wastage’ (see section iii(c)) is an overarching explanation, which occurs when a resource is consumed (in the broad sense) which is not critical for growth, which can inflate interspecific competition relative to intraspecific competition (figure 4). In order for wastage to confer alternative stable states, both species must waste each other's resources. For example, a wild pig digging through soil to find food might destroy habitat for groundhogs; groundhogs, in the process of digging their tunnels, may destroy food resources for wild pigs. If wild pigs are more abundant than groundhogs, groundhogs are at a disadvantage and vice versa. Although examples are most easily conceived of for non-substitutable resources, where each species requires something very different from the environment, the same reasoning should apply to substitutable resources if resource types differ in how nutritious they are to different consumers (i.e., $w_{ik} \neq w_{jk}$). As one might imagine, the probability of two species who reciprocally waste each other's resources coming into contact is likely low (though they may be common in some systems, for example, microbial systems with waste by-products), explaining why positive frequency dependence tends to be a rare occurrence in coexistence studies relative to other competitive dynamics (Buche et al. 2022).

Funding

Funding This project was funded by an NSERC Discovery Grant to R. M. G. (2019-04872) and a BC Graduate Scholarship to J.S.

References

- Abrams, P. A. (2022). *Competition theory in ecology*. Oxford University Press.
- Abrams, P. A., Rueffler, C., and Kim, G. (2008). Determinants of the strength of disruptive and/or divergent selection arising from resource competition. *Evolution*, 62(7):1571–1586.
- Ackermann, M. and Doebeli, M. (2004). Evolution of niche width and adaptive diversification. *Evolution*, 58(12):2599–2612.
- Aoyama, L., Shoemaker, L. G., Gilbert, B., Collinge, S. K., Faist, A. M., Shackelford, N., Temperton, V. M., Barabás, G., Larios, L., Ladouceur, E., Godoy, O., Bowler, C., and Hallett, L. M. (2022). Application of modern coexistence theory to rare plant restoration provides early indication of restoration trajectories. *Ecol. Appl.*, 32(7):e2649.
- Araújo, M. S., Bolnick, D. I., and Layman, C. A. (2011). The ecological causes of individual specialisation. *Ecol. Lett.*, 14(9):948–958.
- Arendt, J. D. (1997). Adaptive intrinsic growth rates: an integration across taxa. *Q. Rev. Biol.*
- Ayala, F. J., Gilpin, M. E., and Ehrenfeld, J. G. (1973). Competition between species: theoretical models and experimental tests. *Theor. Popul. Biol.*, 4(3):331–356.
- Barabás, G., D’Andrea, R., and Stump, S. M. (2018). Chesson’s coexistence theory. *Ecol. Monogr.*, 88(3):277–303.
- Beardsell, A., Gravel, D., Berteaux, D., Gauthier, G., Clermont, J., Careau, V., Lecomte, N., Juhasz, C.-C., Royer-Boutin, P., and Bêty, J. (2021). Derivation of predator functional responses using a mechanistic approach in a natural system. *Frontiers in Ecology and Evolution*, 9.
- Bolker, B. M. (2008). *Ecological models and data in R*. Princeton University Press.
- Brodersen, J., Post, D. M., and Seehausen, O. (2018). Upward adaptive radiation cascades: predator diversification induced by prey diversification. *Trends Ecol. Evol.*, 33(1):59–70.
- Broekman, M. J. E., Muller-Landau, H. C., Visser, M. D., Jongejans, E., Wright, S. J., and de Kroon, H. (2019). Signs of stabilisation and stable coexistence. *Ecol. Lett.*
- Brown, J. H., Gillooly, J. F., Allen, A. P., Savage, V. M., and West, G. B. (2004). Toward a metabolic theory of ecology. *Ecology*, 85(7):1771–1789.
- Buche, L., Spaak, J. W., Jarillo, J., and De Laender, F. (2022). Niche differences, not fitness differences, explain predicted coexistence across ecological groups. *J. Ecol.*, 110(11):2785–2796.
- Carscadden, K. A., Emery, N. C., Arnillas, C. A., Cadotte, M. W., Afkhami, M. E., Gravel, D., Livingstone, S. W., and Wiens, J. J. (2020). Niche breadth: Causes and consequences for ecology, evolution, and conservation. *Q. Rev. Biol.*, 95(3):179–214.
- Chesson, P. (1990). MacArthur’s consumer-resource model. *Theor. Popul. Biol.*, 37(1):26–38.
- Chesson, P. (2020). MacArthur (1970) and mechanistic coexistence theory. *Theor. Popul. Biol.*, 133:13–14.
- Chesson, P. and Kuang, J. J. (2008). The interaction between predation and competition. *Nature*, 456(7219):235–238.
- Fronhofer, E. A., Govaert, L., O’Connor, M. I., Schreiber, S. J., and Altermatt, F. (2018). The shape of density dependence and the relationship between population growth, intraspecific competition and equilibrium population density.
- Gatto, M. (1990). A general minimum principle for competing populations: Some ecological and evolutionary consequences. *Theor. Popul. Biol.*, 37(3):369–388.
- Germain, R. M., Kowalski, J. J., Leonard, S. J., Ou, W. J.-A., Sakarchi, J., and Waters, M. T. (2022). Coexistence in ecology: A mechanistic perspective. monographs in population biology by mark a. McPeck. *Q. Rev. Biol.*, 97(3):222–223.
- Germain, R. M., Matthews, B., and Harmon, L. (2024). Niche theory as an underutilized resource for the study of adaptive radiations. *Cold Spring Harb. Perspect. Biol.*
- Germain, R. M., Mayfield, M. M., and Gilbert, B. (2018). The ‘filtering’ metaphor revisited: competition and environment jointly structure invasibility and coexistence. *Biol. Lett.*, 14(8).
- Germain, R. M., Srivastava, D., and Angert, A. L. (2020). Evolution of an inferior competitor increases resistance to biological invasion. *Nat Ecol Evol*, 4(3):419–425.
- Ghedini, G., Loreau, M., White, C. R., and Marshall, D. J. (2018). Testing MacArthur’s minimisation principle: do communities minimise energy wastage during succession? *Ecol. Lett.*, 21(8):1182–1190.
- Goldberg, D. E. (1990). Components of resource competition in plant communities. *Perspectives on plant competition*, pages 27–49.
- Grainger, T. N., Letten, A. D., Gilbert, B., and Fukami, T. (2019a). Applying modern coexistence theory to priority effects. *Proc. Natl. Acad. Sci. U. S. A.*, 116(13):6205–6210.

- 874 Grainger, T. N., Levine, J. M., and Gilbert, B. (2019b). The 922
875 invasion criterion: A common currency for ecological re- 923
876 search. *Trends Ecol. Evol.*, 34(10):925–935. 924
- 877 Grainger, T. N., Senthilnathan, A., Ke, P.-J., Barbour, M. A., 925
878 Jones, N. T., DeLong, J. P., Otto, S. P., O'Connor, M. I., 926
879 Coblenz, K. E., Goel, N., Sakarchi, J., Szojka, M. C., 927
880 Levine, J. M., and Germain, R. M. (2022). An empiricist's 928
881 guide to using ecological theory. *Am. Nat.*, 199(1):1–20. 929
- 882 Hart, S. P., Freckleton, R. P., and Levine, J. M. (2018). How 930
883 to quantify competitive ability. *J. Ecol.*, 106(5):1902– 931
884 1909. 932
- 885 Hernández-García, E., López, C., Pigolotti, S., and Andersen, 933
886 K. H. (2009). Species competition: coexistence, exclusion 934
887 and clustering. *Philos. Trans. A Math. Phys. Eng. Sci.*, 935
888 367(1901):3183–3195. 936
- 889 Holt, R. D. (1977). Predation, apparent competition, and 937
890 the structure of prey communities. *Theor. Popul. Biol.*, 938
891 12(2):197–129. 939
- 892 Holt, R. D. (2009). Bringing the hutchinsonian niche into 940
893 the 21st century: ecological and evolutionary perspec- 941
894 tives. *Proc. Natl. Acad. Sci.*, 106 Suppl 2(Suppl 2):19659– 942
895 19665. 943
- 896 Hutchings, J. A. (2021). *A Primer of life histories: ecology,* 943
897 *evolution, and application.* Oxford University Press. 944
- 898 Hutchinson, G. E. (1957). Concluding remarks. *Cold Spring* 945
899 *Harb. Symp. Quant. Biol.*, 22:415–427. 946
- 900 Hutchinson, G. E. (1978). *An introduction to population* 947
901 *ecology.* 948
- 902 Jeffries, M. J. and Lawton, J. H. (1984). Enemy free space 949
903 and the structure of ecological communities. *Biol. J. Linn.* 950
904 *Soc. Lond.*, 23(4):269–286. 951
- 905 Johnson, C. A. (2021). How mutualisms influence the coex- 952
906 istence of competing species. *Ecology*, 102(6):e03346. 953
- 907 Johnson, E. C. and Hastings, A. (2022). Towards a heuris- 954
908 tic understanding of the storage effect. *Ecol. Lett.*, 955
909 25(11):2347–2358. 956
- 910 Jolliffe, P. A. (2000). The replacement series. *J. Ecol.*, 957
911 88(3):371–385. 958
- 912 Ke, P.-J. and Letten, A. D. (2018). Coexistence theory and the 959
913 frequency-dependence of priority effects. *Nat Ecol Evol*, 960
914 2(11):1691–1695. 961
- 915 Kraft, N. J. B., Godoy, O., and Levine, J. M. (2015). Plant 962
916 functional traits and the multidimensional nature of species 963
917 coexistence. *Proc. Natl. Acad. Sci. U. S. A.*, 112(3):797– 964
918 802. 965
- 919 Kuno, E. (1991). Some strange properties of the logistic 965
920 equation defined with r and K : Inherent defects or arti- 966
921 facts? *Popul. Ecol.*, 33(1):33–39. 967
- Lanuza, J. B., Bartomeus, I., and Godoy, O. (2018).
Opposing effects of floral visitors and soil conditions
on the determinants of competitive outcomes maintain
species diversity in heterogeneous landscapes. *Ecol. Lett.*,
21(6):865–874.
- Lawlor, L. R. (1980). Overlap, similarity, and competition
coefficients. *Ecology*.
- Leibold, M. A. (1995). The niche concept revisited:
Mechanistic models and community context. *Ecology*,
76(5):1371–1382.
- Leibold, M. A. and Chase, J. M. (2017). *Metacommunity*
Ecology, Volume 59. Princeton University Press.
- Leimar, O., Sasaki, A., Doebeli, M., and Dieckmann, U.
(2013). Limiting similarity, species packing, and the shape
of competition kernels. *J. Theor. Biol.*, 339:3–13.
- Letten, A. D., Ke, P.-J., and Fukami, T. (2017). Linking
modern coexistence theory and contemporary niche theo-
ry. *Ecol. Monogr.*, 87(2):161–177.
- Levins, R. (1968). *Evolution in changing environments:*
some theoretical explorations. Princeton University Press.
- Lincoln, A. E. and Quinn, T. P. (2018). Optimal foraging or
surplus killing: selective consumption and discarding of
salmon by brown bears. *Behav. Ecol.*, 30(1):202–212.
- Loreau, M. (1995). Consumers as maximizers of matter and
energy flow in ecosystems. *Am. Nat.*
- MacArthur, R. (1969a). Species packing, and what com-
petition minimizes. *Proc. Natl. Acad. Sci. U. S. A.*,
64(4):1369–1371.
- MacArthur, R. (1970). Species packing and competitive equi-
librium for many species. *Theor. Popul. Biol.*, 1(1):1–11.
- MacArthur, R. and Levins, R. (1967). The limiting similarity,
convergence, and divergence of coexisting species. *Am.*
Nat., 101:377–385.
- MacArthur, R. H. (1962). Some generalized theorems of nat-
ural selection. *Proceedings of the National Academy of*
Sciences, 48(11):1893–1897.
- MacArthur, R. H. (1969b). Patterns of communities in the
tropics. *Biol. J. Linn. Soc. Lond.*, 1(1-2):19–30.
- MacArthur, R. H. (1972). *Geographical ecology: Patterns in*
the distribution of species. Princeton University Press.
- MacArthur, R. H. and Wilson, E. O. (1967). *The theory of*
island biogeography. Princeton University Press.
- Mallet, J. (2012). The struggle for existence: how the no-
tion of carrying capacity, k , obscures the links between
demography, darwinian evolution, and speciation. *Evol.*
Ecol. Res., 14:627–665.

- 968 May, R. M. (1974). On the theory of niche overlap. *Theor.*1015
969 *Popul. Biol.*, 5(3):297–332. 1016
- 970 McPeck, M. A. (2022). *Coexistence in ecology: A mechanis-*1017
971 *tic perspective*. Princeton University Press. 1018
1019
- 972 Miller, T. E., Burns, J. H., Munguia, P., Walters, E. L., Knei-
973 tel, J. M., Richards, P. M., Mouquet, N., and Buckley,1020
974 H. L. (2005). A critical review of twenty years' use of1021
975 the resource-ratio theory. *Am. Nat.*, 165(4):439–448. 1022
1023
- 976 O'Dwyer, J. P. (2018). Whence Lotka-Volterra? *Theor. Ecol.*,
977 11(4):441–452. 1024
1025
- 978 Ou, W. J.-A., Henriques, G. J. B., Senthilnathan, A., Ke, P.-1026
979 J., Grainger, T. N., and Germain, R. M. (2022). Writing
980 accessible theory in ecology and evolution: insights from
981 cognitive load theory. *Bioscience*, 72(3):300–313. 1027
1028
- 982 Pianka, E. R. (1970). On r- and K-Selection. *Am. Nat.*,
983 104(940):592–597. 1029
1030
1031
- 984 Pigolotti, S., López, C., Hernández-García, E., and Andersen,
985 K. H. (2010). How gaussian competition leads to lumpy or
986 uniform species distributions. *Theor. Ecol.*, 3(2):89–96. 1032
1033
1034
- 987 Ponisio, L. C., Valdovinos, F. S., Allhoff, K. T., Gaiarsa,
988 M. P., Barner, A., Guimarães, Jr, P. R., Hembry, D. H.,
989 Morrison, B., and Gillespie, R. (2019). A network per-
990 spective for community assembly. *Front. Ecol. Evol.*, 7. 1035
1036
1037
1038
- 991 Rankin, D. J., Bargum, K., and Kokko, H. (2007). The
992 tragedy of the commons in evolutionary biology. *Trends*
993 *Ecol. Evol.*, 22(12):643–651. 1039
1040
- 994 Ray, J. (2014). *The Rosetta Stone: And the rebirth of Ancient*
995 *Egypt*. Profile Books. 1041
1042
- 996 Reynolds, S. A. and Brassil, C. E. (2013). When can a single-
997 species, density-dependent model capture the dynamics of
998 a consumer-resource system? *J. Theor. Biol.*, 339:70–83. 1043
1044
1045
1046
- 999 Reznick, D., Bryant, M. J., and Bashey, F. (2002). r- and
1000 k-selection revisited: the role of population regulation in
1001 life-history evolution. *Ecology*, 83(6):1509–1520. 1047
1048
1049
- 1002 Roff, D. (1993). *Evolution of life histories: theory and anal-*
1003 *ysis*. Springer Science & Business Media. 1050
1051
- 1004 Roughgarden, J. (1971). Density-dependent natural selec-
1005 tion. *Ecology*, 52(3):453–468. 1052
1053
- 1006 Roughgarden, J. (1974). Species packing and the competi-
1007 tion function with illustrations from coral reef fish. *Theor.*
1008 *Popul. Biol.*, 5(2):163–186. 1054
1055
1056
1057
- 1009 Roughgarden, J. and Feldman, M. (1975). Species packing
1010 and predation pressure. *Ecology*, 56(2):489–492. 1058
1059
- 1011 Saavedra, S., Rohr, R. P., Bascompte, J., Godoy, O., Kraft, N.
1012 J. B., and Levine, J. M. (2017). A structural approach for
1013 understanding multispecies coexistence. *Ecol. Monogr.*,
1014 87(3):470–486. 1060
1061
1062
1063
- Sakarchi, J. and Germain, R. M. (2023). The evolution of
competitive ability. *Am. Nat.*, 201(1):1–15.
- Schoener, T. W. (1974). Some methods for calculating com-
petition coefficients from resource-utilization spectra. *Am.*
Nat., 108(961):332–340.
- Shoemaker, L. G., Walter, J. A., Gherardi, L. A., DeSiervo,
M. H., and Wisnoski, N. I. (2021). Writing mathemati-
cal ecology: A guide for authors and readers. *Ecosphere*,
12(8).
- Song, C., Barabás, G., and Saavedra, S. (2019). On the con-
sequences of the interdependence of stabilizing and equal-
izing mechanisms. *Am. Nat.*, 194(5):627–639.
- Srinivasan, B. (2022). A guide to the Michaelis–Menten
equation: steady state and beyond. *FEBS J.*
- Sterner, R. W. and Elser, J. J. (2002). *Ecological stoichiom-*
etry: The biology of elements from molecules to the bio-
sphere. Princeton University Press.
- Taylor, D. R., Aarssen, L. W., and Loehle, C. (1990). On the
relationship between r/k selection and environmental car-
rying capacity: A new habitat templet for plant life history
strategies. *Oikos*, 58(2):239–250.
- Tilman, D. (1980). Resources: A graphical-mechanistic
approach to competition and predation. *Am. Nat.*,
116(3):362–393.
- Tilman, D. (1982). Resource competition and community
structure. *Monogr. Popul. Biol.*, 17:1–296.
- Tilman, G. D. (1984). Plant dominance along an experimen-
tal nutrient gradient. *Ecology*, 65(5):1445–1453.
- Usui, T., Lerner, D., Eckert, I., Angert, A. L., Garroway,
C. J., Hargreaves, A., Lancaster, L. T., Lessard, J.-P., Riva,
F., Schmidt, C., van der Burg, K., and Marshall, K. E.
(2023). The evolution of plasticity at geographic range
edges. *Trends Ecol. Evol.*, 38(9):831–842.
- Van Dyke, M. N., Levine, J. M., and Kraft, N. J. B. (2022).
Small rainfall changes drive substantial changes in plant
coexistence. *Nature*, 611(7936):507–511.
- Vandermeer, J. H. and Goldberg, D. E. (2013). *Population*
ecology: First principles – Second Edition. Princeton Uni-
versity Press.
- Vincent, T. L. S., Scheel, D., Brown, J. S., and Vincent, T. L.
(1996). Trade-offs and coexistence in consumer-resource
models: it all depends on what and where you eat. *Am.*
Nat., 148(6):1038–1058.
- Weigelt, A. and Jolliffe, P. (2003). Indices of plant competi-
tion. *J. Ecol.*, 91(5):707–720.
- Yan, X., Levine, J. M., and Kandlikar, G. S. (2022). A
quantitative synthesis of soil microbial effects on plant
species coexistence. *Proc. Natl. Acad. Sci. U. S. A.*,
119(22):e2122088119.

1064 Zou, H.-X. and Rudolf, V. H. W. (2023). Bridging theory
1065 and experiments of priority effects. *Trends Ecol. Evol.*,
1066 38(12):1203–1216.

DRAFT V2

Table 2. Translating MacArthur's consumer-resource model into terms of Lotka-Volterra's competition model

Lotka-Volterra and related quantities	MacArthur	Biological interpretation
A. Consumer intrinsic rate of increase (s_i)	$C_i (\sum_{k=1}^m a_{ik} w_k K_k - T_i)$	s_i is derived from the conversion of consumed energy to new consumers when there are no consumers present, i.e., when the resource is at its carrying capacity (K_k). Under these conditions, energy acquired through consumption ($\sum a_{ik} w_k K_k$) is as high above requirement as is possible in a given environment. Conversion rate C_i determines how energy gains are converted into new consumer individuals. Referred to as " \mathcal{C}_i " in MacArthur (1970).
B. Consumer carrying capacity (L_i)	$\frac{C_i (\sum_{k=1}^m a_{ik} w_k K_k - T_i)}{C_i \sum_{k=1}^m a_{ik}^2 w_k K_k / r_k}$	The numerator is the equation for s_i , whereas the denominator is the unscaled version of a_{ik} (see note in table 2D). Can be simplified in terms of useful production as $\frac{r_k}{a_{ik}} (1 - \frac{R_{ik}^*}{K_k}) = \frac{P_k}{a_{ik}}$ (see table 2G).
C. Interspecific interaction coefficient (α_{ij})	$\frac{C_i \sum_{k=1}^m a_{ik} a_{jk} w_k (K_k / r_k)}{C_i \sum_{k=1}^m a_{ik}^2 w_k (K_k / r_k)}$ $= \frac{U_i U_j}{U_i^2}$	How much energy available ($w_k K_k / r_k$) for growth is lost per individual of consumer i for each individual of consumer j added (numerator), relative to losses due to competition with conspecifics (denominator). Note that this version of Lotka-Volterra quantifies interspecific competition scaled relative to intraspecific competition. This can be undone (as in equation 3) by removing the denominator.
D. Intraspecific interaction coefficient (α_{ii})	$\frac{C_i \sum_{k=1}^m a_{ik}^2 w_k K_k / r_k}{C_i \sum_{k=1}^m a_{ik}^2 w_k K_k / r_k}$ $= \frac{U_i^2}{U_i^2}$ $= 1$	The interpretation is the same as for α_{ij} except now between conspecific individuals. This version of Lotka-Volterra, which MacArthur uses, sets α_{ii} to 1. This can be undone (as in equation 3) by removing the denominator.
E. Resource at consumer equilibrium (R_{ik}^*)	$\frac{T_i}{a_{ik} w_k}$	The resource density when consumer species i and resource species k are in equilibrium. This is as low as the resource concentration will go, and is the point at which consumer energy gain = energy loss. Note: introducing more than one substitutable resource creates conditions for apparent competition among resources and can affect R_{ik}^* in comparison to isolation such that (reformulated from Holt (1977)): $R_{ik}^* = \frac{\frac{T}{a_{ik} w_k} - \sum_{l \neq k}^m K_l \times \frac{a_{il} w_l}{a_{ik} w_k} \times \left(1 - \frac{a_{il} / r_l}{a_{ik} / r_k}\right)}{1 + \sum_{l \neq k}^m \frac{a_{il}^2}{a_{ik}^2} \times \frac{w_l K_l / r_l}{w_k K_k / r_k}}$
F. Resource density at any given level of consumers (\hat{R}_k)	$K_k - \sum_{i=1}^n \frac{(K_k a_{ik} X_i)}{r_k}$	Resource density that is in equilibrium with any given density of consumers. This assumes a separation of timescales between resources and consumers.
G. "Useful" production (P_k)	$r_k - \left(\frac{r_k}{K_k}\right) \left(\frac{T}{a w_k}\right)$	Per capita growth rate of the resource at the consumer's R^* , prior to consumption. This amount should equal total consumption by all consumers at their carrying capacity (i.e., $a_{ik} L_i$) so the net difference is 0 (i.e., both consumer and resource are in equilibrium). Note: this assumes that all consumers have equivalent a_{ik} and T_i , hence, no consumer subscripts. We can simplify this to $r_k (1 - R_k^* / K_k)$ given that $R_{ik}^* = T_i / (a_{ik} w_k)$ (see table 2E).



1067

1068 *Photo of the Rosetta Stone, also known as the "Stone of Rashid" (حجر رشيد) in Arabic, as it is currently held in the British Museum. © The Trustees of the British Museum*

DRAFT V2