

1 **Article Title:** Incorporating traits into consumer-resource models for a mechanistic trait ecology

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22

23 **Abstract**

24 Traits provide a powerful way to infer community assembly processes, responses to environmental  
25 perturbations, and coexistence mechanisms, although most work has focused on plants and has not  
26 incorporated the role of trophic interactions. Here, we briefly review the main goals of trait-based  
27 ecology and highlight recent examples that use traits to study insect communities, specifically  
28 focusing on multitrophic communities and interactions between plants and insect herbivores. We  
29 then introduce a mechanistic multi-species consumer-resource model that incorporates traits  
30 involved in trophic interactions (i.e., leaf toughness and mandible strength) to outline mechanistic  
31 hypotheses for how interactions structure communities. Models show that different ways in which  
32 plants and herbivores interact through their traits, trait-matching vs. strong mandibles,  
33 substantially changes community coexistence patterns. Furthermore, we simulate interactions  
34 across a productivity gradient to generate predictions for how community composition changes  
35 across gradients, given a mechanistic hypothesis. Finally, we evaluate these patterns through the  
36 lens of community trait metrics and Modern Coexistence Theory. The modeling output highlights  
37 that, even under one mechanistic structure (i.e., trait matching between plants and herbivores),  
38 community composition patterns and the relative strength of equalizing and stabilizing  
39 mechanisms change strongly across environmental gradients. Our review provides an avenue for  
40 future research that combines functional traits and mechanistic models to provide a richer  
41 understanding of the processes that structure and maintain diversity in insect communities.

42

43 **Highlights (3-5)**

- 44 • Traits can be used to infer how trophic interactions shape communities
- 45 • Mechanistic models parameterized with trait information predict community patterns
- 46 • Relative strength of equalizing and stabilizing mechanisms changes across gradients

47

48 **Introduction**

49 Understanding the factors that drive patterns of species diversity and community assembly  
50 across space is a central goal in ecology. Traditionally, research has focused on how community  
51 composition of a single trophic level (i.e., taxonomic richness or diversity) changes across  
52 environmental gradients [1,2]. However, it has been established that environmental variation can  
53 simultaneously affect multiple trophic levels, for example, herbivore diversity can be affected  
54 either directly by abiotic factors like climate or indirectly via changes in plant diversity [3,4].  
55 Studies of diversity patterns across trophic levels have typically found positive relationships,  
56 where diversity at one trophic level is correlated with diversity at an adjacent level [5,6]. While it  
57 is generally recognized that taxonomic diversity can be linked to resource differentiation, the  
58 processes that mechanistically link co-variation in structure across trophic levels remain poorly  
59 understood.

60

61 **Functional traits to infer assembly patterns**

62 Trait-based ecology is well suited to elucidate mechanisms shaping communities at more  
63 than one trophic level because it focuses on how functional properties beyond their taxonomic  
64 identities shape interactions with the abiotic and biotic environment [7,8]. Functional traits are  
65 broadly defined as characteristics (morphological, physiological, or phenological) of an organism  
66 that influence its performance and fitness [9]. Over the last two decades, trait-based approaches  
67 have successfully revealed mechanisms that structure plant communities [7,10,11]. For example,  
68 functional traits have been used to show that plant communities shift from slower growing species

69 with conservative resource-use strategies to species characterized by rapid growth and resource  
70 acquisition along gradients of increasing resource availability [12,13]. Traits associated with plant  
71 resource acquisition strategies also alter plant susceptibility to herbivores as articulated in the  
72 resource availability hypothesis from plant defense theory [14]. Slow growing plants occurring in  
73 low-resource environments tend to be better defended and less palatable to herbivores, whereas  
74 faster growing plants in high-resource environments tend to have few defenses and be more  
75 palatable to herbivores [14,15]. However, most functional trait research to date has focused  
76 primarily on plants, and it has only recently been applied to community processes at other trophic  
77 levels [16,17]. Very few studies have investigated functional traits related to plant-herbivore  
78 trophic interactions and how they might contribute to structuring communities [18–21] despite the  
79 potential to identify which plant or herbivore traits may lead to shifts in community composition.

80 One popular use of functional traits is to inference drivers of community assembly, specifically  
81 differentiating between environmental filtering and competition [22]. Communities strongly  
82 structured by inter-specific competition would exhibit trait overdispersion as a result of niche  
83 differentiation, making them more likely to coexist, whereas species with similar traits would most  
84 likely outcome one another. In contrast, environmental filtering is expected to select strongly for  
85 certain functional traits that are necessary for establishment and persistence within a community,  
86 resulting in trait underdispersion, or clustering. Interpreting trait dispersion patterns from  
87 observational data, particularly environmental filtering [23], is difficult unless combined with two  
88 additional steps [16,24]. First, the degree of clustering/dispersion must be shown to correlate with  
89 the hypothesized environmental driver. This step would show that the environmental variable is  
90 the likely driver of trait dispersion patterns. The final step is relating functional traits to abundance  
91 or performance across an environmental gradient [16,24]. This last point connects traits directly to  
92 environmental drivers. For insects, grasshoppers with long wings (i.e., greater dispersal potential)  
93 tended to be more abundant after prescribed fires as indicated by greater community weighted  
94 mean wing length, suggesting that dispersal ability is important for structuring communities  
95 responding to recent fire [21]. In contrast, if competition structures communities, trait patterns  
96 would not correlate with environmental variables. Previous studies have documented correlations  
97 between plant variables (i.e., richness, habitat type) and insect herbivore communities. Therefore,  
98 established theory coupled with functional traits is well poised to address important and  
99 unanswered questions regarding how trophic interactions structure communities.

100

### 101 **Trait-matching to explain assembly of multi-trophic communities**

102 Recent observational trait-based studies of insect communities have demonstrated the roles of  
103 species interactions in structuring community composition (e.g., competition, pollination,  
104 predation [25,26]), identifying how insect communities response to environmental change [27–  
105 30], and how herbivores impact ecosystem functioning [31,32]. Additionally, traits may provide  
106 insight on functional linkages between plants and consumers (i.e., trait matching; [33]). “Trait  
107 matching” refers to traits that specifically enable trophic interactions among species [33]. The most  
108 well-known examples include plant-pollinator interactions (matching proboscis length with  
109 corolla depth; [34]. When there is a mismatch in traits, such as proboscides or bills that are shorter  
110 than flower tubes, this creates an interaction barrier [35]. However, trait matching is also important  
111 for plant-herbivore interactions and can provide insights on how insect communities are structured  
112 [36] as well as how insects influence plant communities and ecosystem function. For example,  
113 plant communities dominated by species with tough leaves tend to support grasshopper species  
114 with strong mandibles [36]. In response, herbivore communities dominated by large-bodied

115 species or species with powerful mandibles can have larger impacts on plant communities than  
 116 small-bodied, weak-jawed species [18,19]. Thus, linking plant and herbivore traits is likely to  
 117 provide mechanistic information beyond the taxonomic diversity approaches of most previous  
 118 studies (e.g., correlating plant and consumer richness). While there is robust and comprehensive  
 119 framework for describing and identifying “trait matching” [33], there is still a lack of model-based  
 120 predictions for how trait matching through plant-herbivore interactions may influence structure of  
 121 communities at both trophic levels.

122

123 **Building functional traits into mechanistic models**

124 To provide general predictions for how functional traits may influence community patterns, we  
 125 modified the classic MacArthur consumer-resource model [37–39] that describes the dynamics of  
 126 a plant–herbivore system. This mechanistic model has biologically informed parameters compared  
 127 to phenomenological models, such as Lotka-Volterra, and therefore is desirable for its  
 128 interpretability [38,40,41]. In this model, plant populations grow at a rate of  $R_i$  and are regulated  
 129 by density-dependence (i.e., intra-specific competition;  $C_i$ ; note that interspecific competition  
 130 coefficients can also be added:  $C_{ik}$ ), as well as by being consumed by herbivores. Herbivore  
 131 populations are regulated by mortality ( $M_j$ ) and the availability of their preferred host plants, which  
 132 they forage on in a density (or frequency) dependent manner (i.e., Type I or II functional response).  
 133 Herbivores do not compete directly, but rather indirectly by suppressing resources [38,41].  
 134 Parameter values are taken from McPeck ([39], pg. 116) with slight modifications (Table 1). For  
 135 simplicity, several model parameters are set to zero (i.e., inter-specific plant competition meaning  
 136 plant species only compete through shared herbivores via apparent competition [42] and handling  
 137 time which specifies a Type I functional response).

138 The basic model can be formally written as:

139 *Plant dynamics:*  $\frac{dP_i}{dt} = P_i \left[ R_i - \sum_k C_{ik} P_k - \sum_j \frac{A_{ij}H_j}{1 + \sum_k A_{ik}L_{kj}H_j} \right]$

140 *Herbivore dynamics:*  $\frac{dH_j}{dt} = H_j \left[ \frac{\sum_i E_{ij}A_{ij}P_i}{1 + \sum_i A_{ij}L_{ij}P_i} - M_j \right]$

141

142 **Table 1.** Model parameters descriptions. Interactions are denoted by matrices ( $ik$  for plant-plant or  
 143  $ij$  for plant-herbivore).

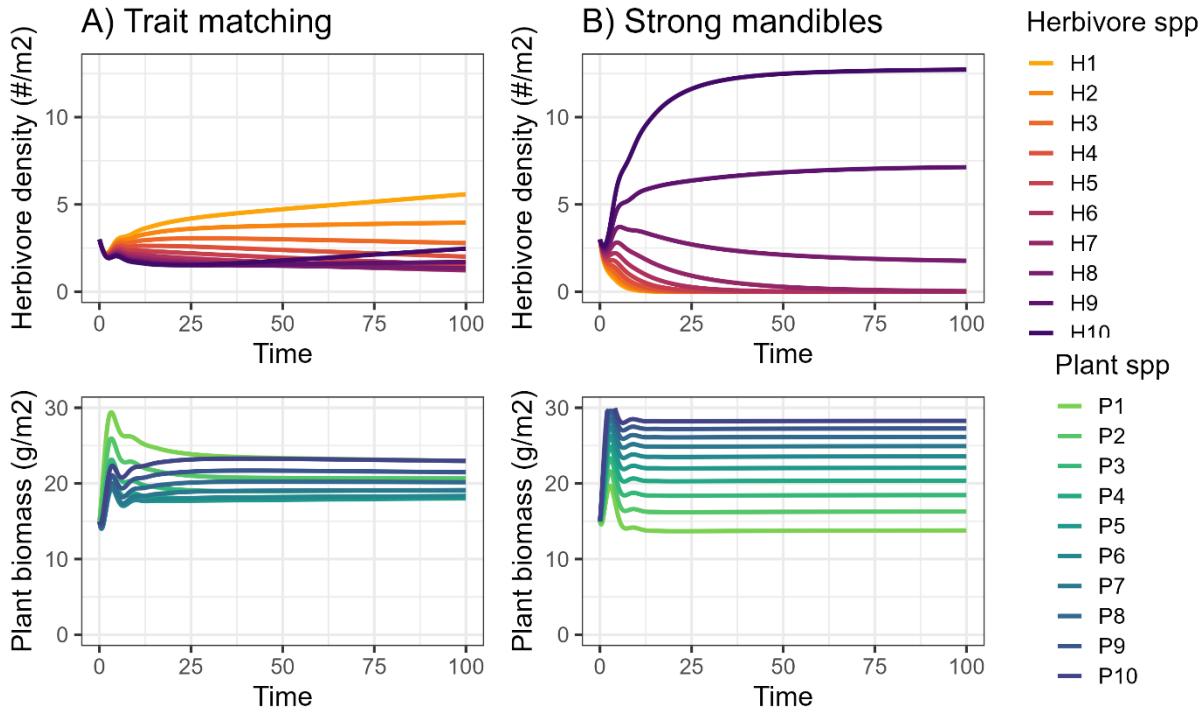
Parameter	Description	Units	Trait(s)	Value(s)
$P_i$	Plant biomass	$\text{g m}^{-2}$		
$H_i$	Herbivore density	$\# \text{ m}^{-2}$		
$R_i$	Intrinsic growth rate of plant $i$	$\text{g m}^{-2} \text{ day}^{-1}$	Could add traits	2
$C_{ik}$	Competition between plant $i$ and $k$	$\text{m}^2 \text{ g}^{-1}$	Could add traits	0.04 (intra), 0 (inter)
$A_{ij}$	Attack rate of herbivore $j$ on plant $i$	$\text{m}^2 \#^{-1} \text{ day}^{-1}$	Mandible strength~ Leaf toughness	max = 0.4; calculated via trait matching
$E_{ij}$	Conversion efficiency of plant $i$ to herbivore $j$	$\text{g g}^{-1}$	Could link traits	0.1
$L_{ij}$	Handling time modifier	$\text{day g}^{-1}$	Could link traits	0
$M_j$	Herbivore mortality rate	$\text{day}^{-1}$	Could add traits	1

144 These mechanistic models can also incorporate plant and herbivore functional traits by allowing  
 145 trait values to modify the attack rates or other parameters of herbivores on different plant species

146 [43]. Previous examples have examined how herbivores with different body sizes can coexist by  
147 linking body size to conversion efficiencies and encounter rates [44]. Feng and DeAngelis [43]  
148 incorporated toxicity into the Type II functional response in a two-plant, one-herbivore system.

149 We incorporated traits by allowing the attack rate matrix,  $A_{ij}$ , to change as a function of the trait  
150 similarity (normalized Gaussian-shaped function, similar to [33]) between mandible strength and  
151 leaf toughness (“Trait matching hypothesis”):  $A_{ij} = \exp\left(-\beta \left(\frac{m_j/\bar{m}}{t_i/\bar{t}} - 1\right)^2\right)$  where  $m_j$  and  $t_i$  are  
152 mandible strength and leaf LDMC, respectively, for herbivore  $j$  and plant  $i$ , and  $\beta$  controls the  
153 spread of the trait matching ( $\beta = 10$  in this case), columns are then normalized so that  $\sum A_{ij} = 0.5$   
154 (Appendix A, Fig. A1A). Trait matching between mandible strength and leaf toughness  
155 (specifically, leaf dry matter content, LDMC) has been empirically linked in several previous  
156 studies with plants and grasshoppers [19,21,36,45]. We used hypothetical values based on the  
157 ranges of LDMC and mandible strength from Grosdidier et al. [21]. An alternative hypothesis is  
158 that stronger mandibles are always better, which we incorporated into a second model (“Strong  
159 mandibles hypothesis”), assuming that weak-mandible species would experience a “trait-barrier”  
160 when trying to consume tough leaved plants [33]. We used a monotonic relationship such that  
161 stronger mandible herbivores had greater attack rates across the board and low LDMC are  
162 preferred by all herbivores:  $A_{ij} = \alpha_0 * \frac{m_j/\bar{m}}{t_i/\bar{t}}$  where  $\alpha_0$  is a base attack rate (0.04; Fig. A1B). Both  
163 models contained 10 plant and herbivore species, with increasing mandible strengths (H1 <...<  
164 H10) and leaf toughness (i.e., LDMC, P1 <...< P10). All other parameters were as in Table 1. We  
165 focus specifically on attack rate,  $A_{ij}$ , because these traits should impact the probability of an  
166 herbivore consuming a plant, although other traits could be added to affect other model parameters,  
167 based on the mechanism (Table 1).

168 The results of our model showed that the two most abundant herbivore species in the “trait  
169 matching” model had the most dissimilar traits (H1 and H10) and all 10 species persisted through  
170 100 time points (Fig. 1A). The two most abundant species in the “strong mandibles” model, not  
171 surprisingly, had the strongest mandibles (H10 and H9) and most other species went extinct (Fig.  
172 1B).



173  
 174 **Fig. 1.** A) Abundances of 10 herbivore (top) and 10 plant (bottom) species over 100 iterations assuming  
 175 trait-matching between leaf toughness and mandible strength based on normalized (%) similarity of  
 176 traits and B) assuming stronger mandibles are better for consuming plants. Traits of the plant and  
 177 herbivore species are ordered from low (P1, H1) to high (P10, H10).

178

179 **Mechanistic models to predict functional trait dispersion across gradients**

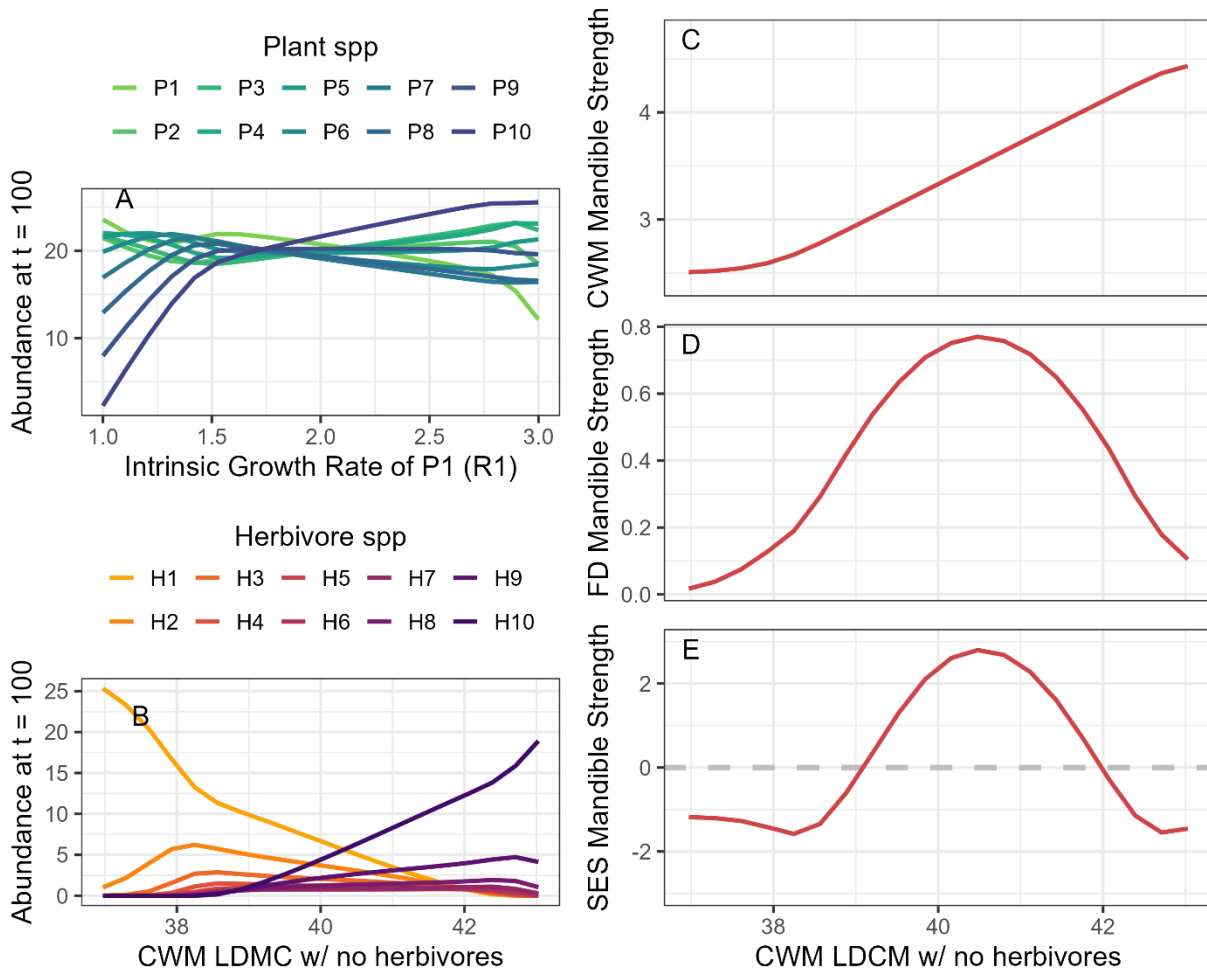
180 We further modified the above models to develop alternative hypotheses for how interactions  
 181 should influence communities across environmental gradients. We simulated an environmental  
 182 gradient of plant community LDMC by varying the growth rates of all plant species systematically  
 183 across 20 growth rate values. In these simulations, average growth rate of the plant community  
 184 was held constant at  $R = 2$ , such that if  $R1 = 1$  then  $R10 = 3$ , and so forth (see McPeck [39], pg.  
 185 100-118). Without herbivores present, this creates a gradient from low to high community  
 186 weighted LDMC (Appendix B).

187 Next, we added herbivores, assuming the “trait matching” scenario described above. In this  
 188 scenario, the plant community follows the same general pattern of increasing community LDMC,  
 189 although herbivory does dampen this pattern to some degree (Fig. 2A). Herbivore abundance also  
 190 changes substantially across this gradient, with the weakest-mandibled species strongly decreasing  
 191 and the strongest-mandibled species increasing, as expected, with all species coexisting at  
 192 intermediate levels of the gradient (Fig. 2B).

193 At the community level, weighted mean (CWM) mandible strength of the herbivore community  
 194 increased along the LDMC gradient (Fig. 2C). Interestingly, mandible trait diversity, measured as  
 195 Rao’s entropy, peaked at the center of the LDMC gradient, likely because plant community traits  
 196 were most even, and therefore provided even and diverse resources for the herbivores. Although  
 197 the mechanism structuring herbivore communities is bottom-up environmental filtering, driven by  
 198 the traits of the plant community, the standardized effect sizes shifted strongly across the LDMC

199 gradient (Fig. 2E). However, at low and high CWM LDMC values, herbivore mandible traits were  
 200 strongly underdispersed, which appropriately detected the environmental filtering. However, at  
 201 intermediate values of CWM LDMC, herbivore mandible traits were strongly overdispersed,  
 202 which is typically an indicator of strong competition. However, in this case, the trait overdispersion  
 203 resulted from the high evenness of the plant traits (Fig. 2A). Although herbivores are competing  
 204 indirectly for resources by depressing plant abundance (compared Fig. 2A to Fig. B1 in Appendix  
 205 B), herbivore species are only excluded at the extremes of the LDMC gradient, where trait  
 206 matching is most intense.

207



208 **Fig. 2.** A) Abundances of 10 plant species simulated across 20 different growth rates that were  
 209 systematically varied across all 20 plant species. Herbivores are present in this model, depressing plant  
 210 abundances (see Appendix for no-herbivore model). Shown on the x-axis is the growth rate of plant  
 211 species 1 ( $R_1$ ). B) Abundances of 10 herbivore species across the simulated gradient of plant LDMC.  
 212 Responses of herbivore community mandible metrics across the LDMC gradient are shown, C)  
 213 community weight mean, D) Rao's entry as a measure of functional dispersion, and E) standardized  
 214 effect sizes, where negative values represent trait underdispersion and positive values represent trait  
 215 overdispersion. x-axis in B-E is the CWM LDMC from the no-herbivore model in Appendix.  
 216

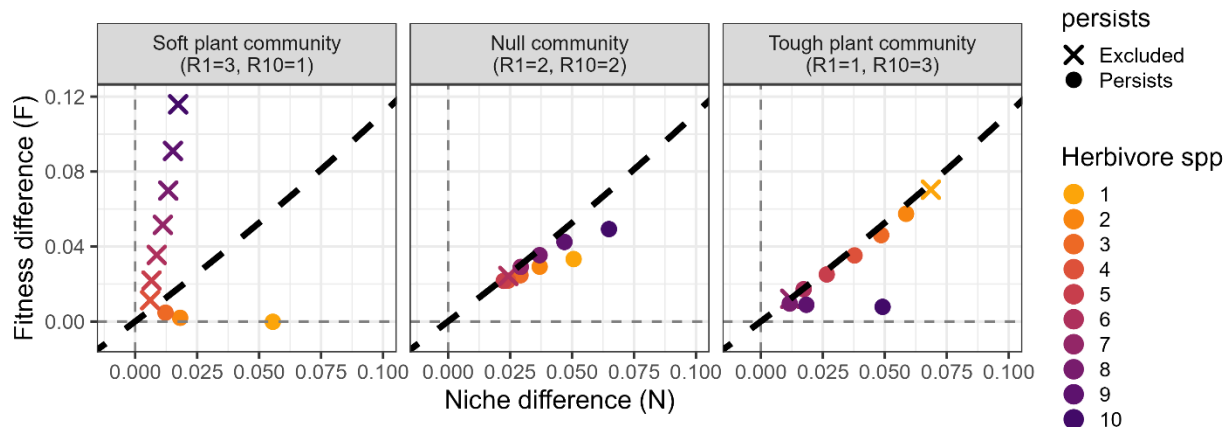
217

218 **Linking to modern coexistence theory**

219 Modern coexistence theory centers on stabilizing and equalizing mechanisms to explain how  
 220 species coexist in diverse communities despite competing for shared resources [46]. Stabilizing  
 221 mechanisms, or niche differences, occur when species possess different traits that allow them to  
 222 exploit different resources, thereby minimizing interspecific competition relative to intraspecific  
 223 competition. Equalizing mechanisms, or fitness differences, reduce competitive disparities by  
 224 minimizing differences in intrinsic growth rates or fecundities, preventing consistent advantages  
 225 of one species [47]. Mechanistic model offer a powerful way to connect functional traits to  
 226 coexistence outcomes because they can be parameterized with functional traits data and contain  
 227 the elements necessary to calculate niche and fitness differences, and therefore coexistence criteria  
 228 [48–50].

229 New definitions of niche and fitness differences now account for a broader range of interactions,  
 230 from competition to facilitation, making it possible to test higher-order interactions beyond simple  
 231 pairwise relationships [51–53]. This  $\mathcal{N}$ - $\mathcal{F}$  mapping framework has several advantages over  
 232 traditional metrics, mainly that estimates are made at the species level (i.e., whether it suppresses  
 233 itself more than it is suppressed by the community) rather than only pairwise comparisons and can  
 234 accommodate multiple trophic levels simultaneously [54,55]. For three scenarios across the  
 235 LDMC gradient described above for the trait-matching hypothesis, we estimated niche and fitness  
 236 differences across a range of parameter values. In the “null” community, where all plant species  
 237 have the same growth rate ( $R = 2$ ), niche differences are large enough overcome fitness differences,  
 238 stabilizing coexistence for most species (Figure 3). However, in the “soft” community, the three  
 239 herbivore species that are best matched to the plant community (i.e., H1-H3 with the weakest  
 240 mandibles) persist while the others are excluded. The “tough” community is in between, with most  
 241 species persisting especially the ones best matched to the plant community (e.g., H9-10).

242



243

244

245 **Fig. 3.** Three scenarios with plant communities simulated across a gradient of CWM LDMC assuming  
 246 the “Trait-matching” hypothesis. Points show each herbivore species mapped into Niche and Fitness  
 247 difference space (Spaak et al. 2021). Thick diagonal dashed line is a 1:1 line that indicates  $N = F$ ,  
 248 where a species will be excluded left of line ( $N < F$ ) or persist right of line ( $N > F$ ) (see Spaak et al.  
 249 2021).

250

251 **Future directions**

252 There is growing evidence that ecological assemblages can be interpreted using trait-based  
253 approaches though the lens of stabilizing and equalizing mechanisms. Theoretical models are a  
254 valuable tool for advancing this understanding and can be used to generate predictions across a  
255 range of scenarios, including various trophic levels and abiotic conditions. Mechanistic models in  
256 particular are particularly powerful because they incorporate rich biological information  
257 [38,48,50] and can generate predictions for a range of scenarios, including for multiple traits,  
258 incorporating predators or mutualists [39,56]. They are also highly flexible, and can accommodate  
259 various functional responses that may better capture ecological dynamics with different outcomes  
260 for coexistence [57], including capturing indirect effect like apparent competition [42] as well as  
261 incorporating more traits for multidimensional inference [58].

262 An additional important source of variation that should be considered when generating future  
263 scenarios is intraspecific trait variation, as it represents the breadth of a species' niche and has been  
264 shown to alter stabilizing and equalizing dynamics under different abiotic conditions [59–61]. As  
265 with most trait-based interactions across trophic levels and overall communities, the effects of  
266 individual trait variation on species fitness and competitive advantage remain largely unknown.  
267 Finally, it is critical to consider that, as ecological scenarios become more complex, multiple  
268 processes can result in similar patterns and trait metrics should be chosen carefully to investigate  
269 specific underlying processes [62].

270

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273

#### 274 **Author Contributions**

275 **Philip Hahn:** Conceptualization, Methodology, Formal analysis, Writing – Original Draft,  
276 Visualization, Funding acquisition. **Lucia Navia:** Methodology, Writing – Review and Editing.

277 **Antonia Millet:** Methodology, Writing – Review and Editing. **Robert Holt:** Formal analysis.

278 **Carolina Baruzzi:** Conceptualization, Methodology, Writing – Review and Editing.

279

#### 280 **Data availability statement**

281 No data were used in this article. R code for the models is available on github:

282 [https://github.com/hahnp13/Hahn\\_COIS2026\\_MCR.git](https://github.com/hahnp13/Hahn_COIS2026_MCR.git)

283

284 **Declaration of interests:** none.

285

#### 286 **Declaration of generative AI and AI-assisted technologies in the manuscript preparation 287 process:**

288 During the preparation of this work, the lead author used Claude Sonnet 4.6 ran on the University  
289 of Florida's [NaviGator Chat](#) to generate R code. Prompts were written to be very detailed for  
290 specific purposes. The conversation was built as a retrieval automated generation (RAG) so that  
291 the model had access to published resources and R scripts written by the lead author that were  
292 uploaded into the chat session. All code produced by the model was reviewed and revised by the  
293 lead author.

294

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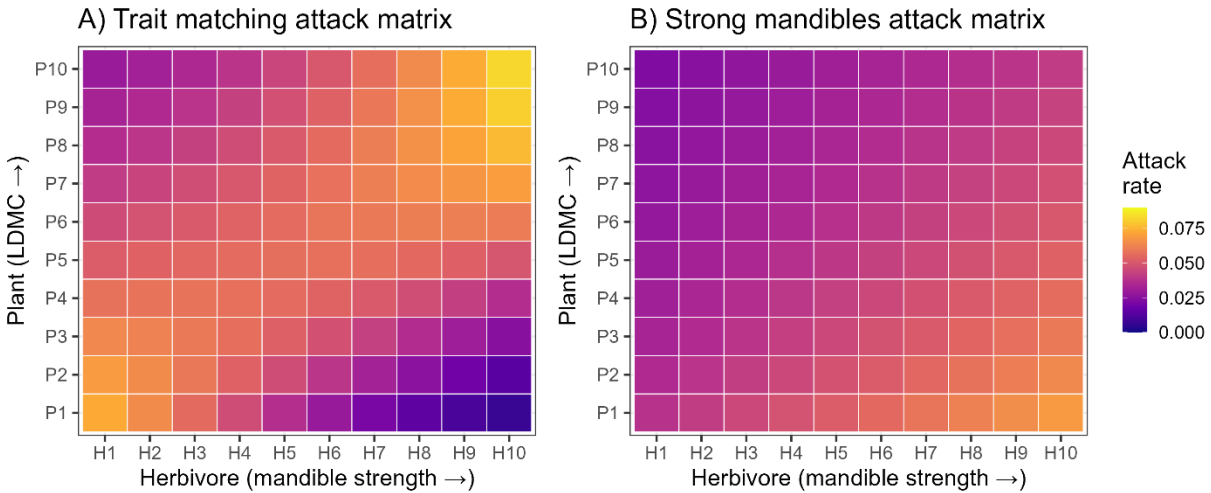
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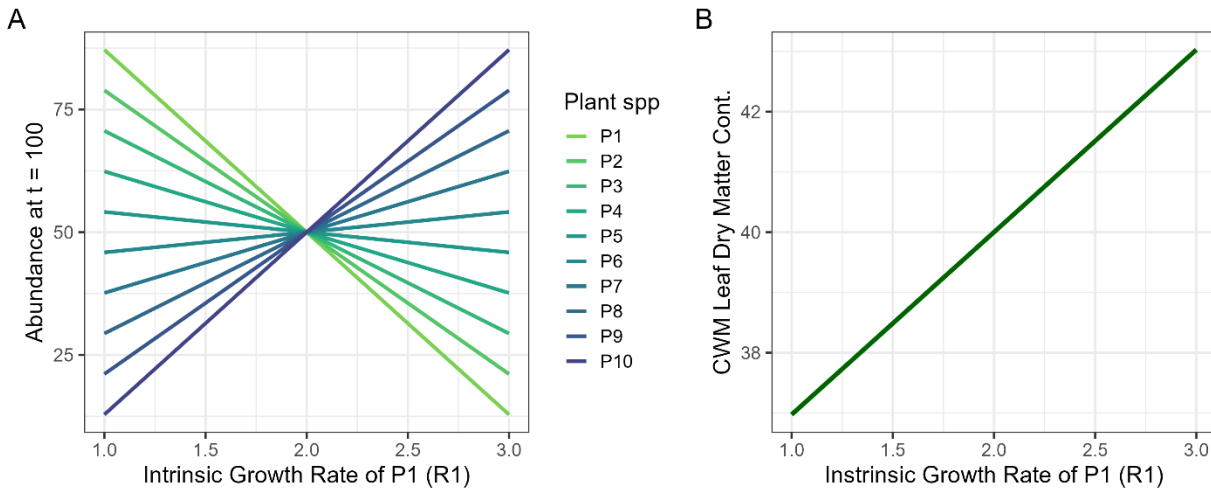
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471 **Appendix A**  
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 475 **Figure A1.** Attack rate matrices for A) trait matching and B) strong mandibles hypotheses.  
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478 **Appendix B**



479  
 480 **Figure B1.** A) Abundances of the 10 plant species simulated across 20 different growth rates that  
 481 systematically varied the growth rates (i.e., productivity) of all species. In these simulations,  
 482 average growth rate of the community was held constant at  $R = 2$ , such that if  $R1 = 1$  then  $R10 =$   
 483  $3$ , and so forth. Shown on the x-axis is the growth rate of plant species 1 ( $R1$ ). Since the plant  
 484 species increase in leaf dry matter content (LDMC), the simulations created B) a gradient of  
 485 community weighted mean (CWM) LDMC. The CWM values were calculated from the  
 486 abundances of the 10 plant species at time 100 for each of  $R$  values. No herbivores are present in  
 487 these simulations.