

1 **A conceptual model for linking traits to plant community assembly using**
2 **experiments and simulations**

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12

13 **Abstract**

14 1. We review the underlying principles of plant community assembly and build a
15 conceptual model into which we map experiments and simulation approaches.

16 2. In this model, environmental filtering selects individuals from a species pool based
17 on non-independent traits bounded by trade-offs. The feedback of communities on
18 environmental factors mimics plant-plant interactions, producing fine-scale heterogeneity
19 and spatial/temporal nesting among factors, which affect trait diversity in the communities.

20 3. Synthetic community experiments focus on the effects of a target species mixture on
21 ecosystem functioning, and weeding non-target species usually halts the assembly process.
22 Experiments on natural communities involve manipulating established assembly processes
23 in pre-existing communities through species removal or addition, altering resources,
24 conditions, or disturbances.

25 4. Stochastic, individual-based models can simulate plant metacommunities, either
26 based on ecophysiological mechanisms or statistical approaches to predict the successful
27 establishment of individual plants based on their traits and local conditions and implicitly
28 model plant interactions through the feedback of the community on the environment.

29 5. *Synthesis*. Experiments and simulation models are promising tools for studying plant
30 community assembly, yet further exploration is needed on the coordination between
31 functional traits during environmental filtering, the feedback from the existing community
32 on environmental factors, and nested environmental factors creating fine-scale
33 heterogeneity.

34 **Keywords:** Biodiversity experiments, Biotic filtering, Community assembly, Community
35 simulation, Environmental filtering, Plant-plant interactions, Traits

37 **Introduction**

38 Ecologists have long been searching for links between plant traits and environmental factors
39 that may explain how plants assemble in communities (Grime, 1979; Keddy, 1992; Pillar &
40 Orłóci, 1993; Raunkiaer, 1934; Warming, 1909). However, revealing the assembly processes
41 that generate observed trait patterns is not straightforward (e.g. Münkemüller et al., 2020).
42 While we often examine trait-environment relations by integrating traits at the community
43 level and examining patterns (Bruehlheide et al., 2018; de Bello, 2021; Pillar et al., 2009), the
44 assembly process selects individuals with sets of non-independent, *coordinated traits* whose

45 relations may be restricted by trade-offs (Díaz et al., 2016; Joswig et al., 2022).
46 Consequently, the filtering effects of environmental and biological factors on a functional
47 trait perspective are often blurred (Anderegg, 2023). Furthermore, in community assembly
48 processes, some traits may become more limiting than others depending on the filtering
49 strength of the factors upon each trait (Pillar et al., 2021). Additionally, ecologists often
50 differentiate between abiotic factors (such as temperature, precipitation, and soil properties)
51 and biotic ones represented by plant-plant interactions (e.g. Bennett & Pärtel, 2017). The
52 effects of abiotic and biotic factors can be distinguished by the convergence and divergence
53 patterns generated at the community level and by examining shifts in trait-based
54 dissimilarities from the species pool to within communities (Bennett & Pärtel, 2017; de
55 Bello et al., 2012). However, environmental factors may change during the community
56 assembly process, not only by external drivers but also due to *feedback* from the existing
57 community (HilleRisLambers et al., 2012). Moreover, environmental factors may be
58 spatially and temporally structured at different scales (Grime, 2002), so their effects are
59 often *nested* and influence community assembly processes at a finer resolution than the grain
60 size of the studied community units.

61 In this review/synthesis, we examine concepts, look into the literature for evidence
62 and build a conceptual model (Figure 1) to tackle these issues by using a range of
63 experiments and simulation models in the search for causal links between traits and
64 community assembly. In a broad sense, we consider a community assembly experiment any
65 experiment involving the manipulation of community components and/or environmental
66 factors used to explore their effects on the community assembly processes. Experiments
67 designed for other aims might also be used for this purpose, such as the well-known
68 biodiversity-ecosystem functioning experiments, that manipulate species diversity and/or
69 composition to examine species loss effects on ecosystem functioning (Bruehlheide et al.,

70 2014; Schmid et al., 2017). These experiments may also manipulate environmental factors
71 (e.g. climatic conditions, resources, disturbances) for evaluating how species are locally
72 selected. Here we focus on those experiments considering plant traits. By manipulating
73 species diversity, community composition, and/or environmental factors that might alter the
74 community composition, we can assess how species (and their traits) relate to each other,
75 affecting the community assembly and ecosystem functions. Such an experimental setting,
76 therefore also allows for exploring spatially nested ecological filters.

77 Plant community assembly is often studied by analysing patterns in community data
78 from both experimental and non-experimental setups. However, while these patterns can
79 suggest assembly processes, the conclusions are rather weak since other implicit processes
80 could also produce similar patterns (Botta-Dukát & Czúcz, 2016). To gain a better
81 understanding of assembly processes, researchers can use community simulation models that
82 integrate information on functional traits in a species pool and their relations with
83 environmental factors and ecosystem effects (e.g. Pillar et al., 2021; Scheiter et al., 2013).
84 Such models can simulate communities and explore assembly processes that normally
85 cannot be accessed through experiments or observational data due to limitations in spatial
86 and temporal resolution or study extent. By comparing the patterns observed in simulated
87 communities with expected patterns based on proposed processes, researchers can
88 empirically verify their hypotheses. However, simulation results depend on the assumptions
89 and parameters chosen by the researcher, and therefore it is essential to carefully review and
90 define how community assembly can be related to traits through simulations.

91 In this paper, we discuss the underlying concepts of a general causal model and the
92 potential and limitations of ecological experiments for studying plant community assembly.
93 Specifically, we focus on experiments in which communities are assessed with regard to
94 their dynamics and functioning after being assembled from scratch by sowing species

95 mixtures, or being (re)assembled by removing or adding target species or manipulating
96 environmental factors in natural communities. We also discuss the possibilities for
97 integrating such conceptual and experimental approaches with predictive computational
98 models that simulate community assembly based on rules representing the proposed
99 processes. This is a non-exhaustive review where for both types of studies, experiments and
100 computational models, we focus only on studies that specifically deal with functional traits.

101

102 **A conceptual model for community assembly**

103 Plant community assembly is a process that takes place in local communities and involves a
104 species pool. In Figure 1, we present a conceptual causal model that we advocate for the
105 study of plant community assembly and ecosystem functioning. Accordingly, species co-
106 occur in communities that are arbitrarily defined at a given spatial/temporal scale. The set of
107 communities composes a metacommunity (Leibold et al., 2004) represented by matrix **C**.
108 Environmental factors (matrix **E**) filter species and this process is mediated by the traits of
109 each species individuals (matrix **T**). The idea of a hierarchy of factors is appealing,
110 represented by decreasing grain sizes of **E**, starting with regional climate and moving to
111 local disturbances, microclimate and soil conditions acting as serial filters. These filters
112 select those species from the species pool that will establish successfully at a community site
113 (HilleRisLambers et al., 2012; Keddy, 1992). However, the fate of an individual plant is
114 determined by all these factors acting simultaneously, independently or interacting with each
115 other in their local selection effects on individuals, given the individuals' trait values
116 (Belyea & Lancaster, 1999). Furthermore, considering the number of communities is
117 sufficiently large, the regional species pool is defined by the set of species occurring in the
118 metacommunity, while the habitat- or site-specific species pool (Bennett & Pärtel, 2021;
119 Zobel, 2016) is defined by the species in the regional species pool that, due to their traits

120 (Bennett & Pärtel, 2017), can potentially occur under the given local environmental
121 conditions in a community site.

122 It is worth remembering that the selected units in the filtering process (Figure 1a) are
123 individual organisms with sets of traits that cannot be separated physically in response to
124 different factors (Violle et al., 2007). While traits vary mostly between species, some traits
125 may have high intraspecific variability (Siefert et al., 2015) that is relevant for community
126 assembly (Carmona et al., 2019; Davrinche & Haider, 2021; Kraft et al., 2014; Pérez-Ramos
127 et al., 2019). Additionally, the traits are not independent of each other, and the multivariate
128 trait space of the regional species pool is not completely filled by the possible combinations
129 of trait values (Díaz et al., 2016; Joswig et al., 2022) due to trade-offs related to ecological
130 and phylogenetic constraints in the evolution of plants (Moles & Westoby, 2006; Wright et
131 al., 2004). As a result, the selection process is mediated by a set of coordinated trait values
132 within the same organism rather than single traits. Grime (1974, 1977) introduced the
133 concept of whole-plant strategies to relate traits with environmental factors (i.e. the CSR
134 model), which may be helpful for understanding community assembly when placed in the
135 context of trade-offs (Pierce et al., 2013).

136 The community level functional descriptor \mathbf{Q} in Figure 1a is defined through trait-
137 based analysis. This approach integrates the taxonomic composition (a matrix \mathbf{C} of species
138 by sites) with the corresponding traits (a species by traits matrix \mathbf{T}) and scales them to the
139 community level (matrix \mathbf{Q}). As \mathbf{T} is carried by the filtered species in \mathbf{C} , which define \mathbf{Q} by
140 computation, it is incorrect to attribute a causal link between these matrices (Grace et al.,
141 2022). For simplicity, \mathbf{Q} is represented as $\mathbf{Q} \sim \mathbf{T}\mathbf{C}$ (or $\mathbf{T}\mathbf{C} \sim \mathbf{Q}$). Depending on the analytical
142 approach, \mathbf{Q} may be a matrix of sites described by community-weighted mean (CWM) traits
143 or some measure of functional diversity (de Bello, 2021). Alternatively, \mathbf{Q} may be a
144 composition matrix of sites described by species after fuzzy-weighting (Pillar's smoothing,

145 Pillar et al., 2009), which is analogous to the Beals transformation (De Cáceres & Legendre,
146 2008). However, instead of co-occurrence probabilities, fuzzy-weighting is based on the trait
147 (or phylogenetic, Duarte et al., 2016) similarities between the species in the regional species
148 pool. That is, a species not currently present has a probability of belonging to a local
149 community given the similarities of this species to the ones that are present in the local
150 community (Pillar et al., 2009; Pillar & Duarte, 2010).

151 Besides environmental filtering, species interactions – such as competition and
152 facilitation – are also considered important processes in community assembly (Chesson,
153 2000; Diamond, 1975; Götzenberger et al., 2012; HilleRisLambers et al., 2012). Most of the
154 research regarding the assembly process has been focused on competitive interactions
155 between species, suggesting interspecific competition leads to nonrandom co-occurrence
156 patterns (Diamond, 1975). The competitive effect refers to the ability of a species to
157 suppress the growth of neighbouring plants mainly through the depletion of resources, while
158 the competitive response refers to the ability of a species to tolerate resource levels that have
159 been reduced due to competition (Goldberg, 1990; Goldberg & Werner, 1983). The
160 relevance of positive interactions in community assembly is also well documented (Bertness
161 & Callaway, 1994; Michalet, 2007; Zhang et al., 2022). Facilitation has been proposed as an
162 important process for community assembly, in which a species modifies resources or
163 conditions with a positive effect on the performance of its neighbours (Callaway & Walker,
164 1997; Michalet & Pugnaire, 2016). The relative strength of facilitative–competitive
165 outcomes have been demonstrated along environmental gradients in different ecosystems
166 (Bertness & Callaway, 1994; He et al., 2013). Furthermore, these interactions drive priority
167 effects in plant community assembly (Fukami et al., 2005; Roscher et al., 2014; Stuble et al.,
168 2017).

169 Nevertheless, in the end plant-plant interactions are mostly mediated by
170 environmental filtering, since biotic interactions are mediated by the effect or the response
171 of the interacting species on resources and/or conditions (Schöb et al., 2017). Functionally
172 translating this concept (Figure 1a), the plant traits integrated at the community level in **Q**
173 (e.g. community-weighted mean, functional diversity, fuzzy-weighted composition) may
174 indirectly affect **E** (e.g. light and nutrient availability, microclimate, disturbances by
175 herbivory and fire) through ecosystem functions or processes in **F** (e.g. productivity, and
176 processes enhancing palatability or flammability). Though testing causal models with
177 feedback is complex and requires advanced techniques (Grace, 2006; Shipley, 2000), such
178 feedback paths have an impact on the community assembly process and should not be
179 ignored. What may be interpreted as a plant-plant interaction may actually be mediated by
180 the feedback (HilleRisLambers et al., 2012) from the existing community on resources and
181 conditions, the “interaction milieu” *sensu* McGill et al. (2006), or by disturbance (e.g. Adler
182 et al., 2001; Grime, 2006), which consequently create fine-scale heterogeneity. In this case,
183 the specificity of plant-plant interactions may be driven by the plant traits in their ecosystem
184 effects and community assembly responses (Schöb et al., 2017), which can be affected by
185 phenotypic plasticity (Pérez-Ramos et al., 2019). Additionally, the environmental
186 heterogeneity created in this process is often nested (Figure 1b), as some environmental
187 factors may be modified at a finer scale than others and remain unmeasured at a finer grain
188 than the observed community units (Pillar et al., 2021; Vellend, 2016). This idea carries
189 important consequences for the simulation and understanding of community assembly, as
190 the relevant process that selects the species in the end is environmental filtering, which
191 occurs at a very fine spatial/temporal scale (Price et al., 2014). Further, as demonstrated in
192 Pillar (2023), when there is interaction between factors in the environmental filtering process
193 based on species traits, the expected trait values at the community level will exhibit a

194 divergence pattern. This occurs because the selection effect driven by one factor is
195 modulated by another factor, resulting in a pattern of beta trait divergence associated with
196 either factor. Since the underlying ecological factors are often spatially nested, beta diversity
197 between communities described at a finer grain size may be observed as alpha diversity
198 within communities at coarser grains. Furthermore, in nature, feedback-driven factors are
199 often hidden and difficult to measure. These hidden factors, although interacting with
200 measurable ones, can generate trait divergence patterns.

201 Many experiments that investigate functional aspects of plant community assembly,
202 especially those that assess feedback processes (e.g. plant-soil feedback), indicate that the
203 assembly process affected by biotic filters can actually be translated as a shift in abiotic
204 conditions/resources at a very fine scale (Helsen et al., 2016; Teixeira et al., 2020). The
205 same idea is theoretically demonstrated by Adler et al. (2013). Therefore, in the context of
206 plant-plant interactions, we question the biotic-abiotic duality as two separate abstractions
207 and suggest that they are not separable; it is only a matter of considering or not the
208 intermediate processes. For example, in the case of competing plants, they deplete resources
209 which subsequently become scarcer for one another. Nurse plants, on the other hand, benefit
210 nearby plants through various mechanisms such as reducing abiotic stress, improving soil
211 moisture and nutrients, and offering protection against herbivores (see e.g. Filazzola &
212 Lortie, 2014). Palatable plants may draw grazers to a feeding spot where the feedback of
213 plant composition on animal behaviour leads to the formation of grazing lawns interspersed
214 by vegetation patches dominated by taller unpalatable plants (Caram et al., 2023; Fischer et
215 al., 2019). Moreover, allelopathic plants release chemical substances that might affect the
216 growth of neighbouring plants (Hierro & Callaway, 2021). This idea is illustrated in Figure
217 1, where we intentionally do not place a direct connection of species composition to itself
218 (C->C). Therefore, in this conceptual model, all biotic community assembly processes occur

219 through the environment, i.e. $Q \rightarrow F \rightarrow E \rightarrow (TC \sim Q)$. In Figure 1b, the thick arrow on the small
220 **E** represents the fine-scale effects of the community on assembly feedback. While thinner
221 arrows pointing to the bigger **E**s represent the decreasing importance of feedback effects on
222 environmental conditions at coarser scales.

223 The idea of tracing the feedback path $Q \rightarrow F \rightarrow E \rightarrow (TC \sim Q)$ to understand plant-plant
224 interactions is appealing. However, measuring the environmental factors that undergo
225 modifications during this process can be challenging, especially at the spatial and temporal
226 resolutions in which changes occur in the plant communities. Nonetheless, Pillar et al.
227 (2021) demonstrated that even when hidden, the environmental factors involved in
228 community assembly can be inferred, as they leave “ghost marks” in the way plants
229 assemble according to their traits, which is reflected in **Q**. The higher the match between the
230 fuzzy-weighted composition in **Q** and the Beals transformed species composition matrix (De
231 Cáceres & Legendre, 2008), derived from species co-occurrence probabilities, the better the
232 traits considered in fuzzy-weighting reflect co-occurrence, i.e., the way species have
233 assembled in the communities (Pillar et al., 2021). Furthermore, with an appropriate
234 sampling design, this analysis could be done at different community unit scales (analogously
235 as suggested by Münkemüller et al., 2020). Once the most relevant traits in community
236 assembly are identified, the underlying environmental factors can be inferred if their
237 relations with the traits are known. For example, specific leaf area is recognized as a trait
238 that responds to nutrient supply. Among these factors, those that portray fine-scale variations
239 within communities are more likely the ones influenced by the community feedback.
240 Therefore, we emphasise the importance of considering this feedback path when designing
241 experiments and simulation models to explore community assembly.

242

243 **How experiments help understanding community assembly**

244 In general, we can find different types of experiments depending on the initial conditions of
245 the target community. These may be synthetic community experiments established from
246 scratch, or natural community experiments in which the communities are manipulated by
247 removals, additions or by changing resources and conditions. However, some studies have a
248 combined experimental approach (e.g. addition experiments in synthetic communities).

249

250 *Synthetic community experiments*

251 Synthetic community experiments are a classic type of biodiversity experiment (e.g.
252 Tilman et al., 2001). However, as we will demonstrate in this section, the utility of synthetic
253 community experiments in addressing questions about community assembly is limited
254 unless the process of community assembly is not hindered by the removal of non-target
255 species. In these experiments, the composition of the species mixtures is defined a priori and
256 then sowed or planted in the field or in pots. To explore the effects of community functional
257 structure on ecosystem processes and community assembly, functional trait information can
258 be integrated with species' taxonomic identity, allowing a posteriori analysis (e.g. Roscher et
259 al., 2014) of the experiment's results.

260 A further deployment of the synthetic community experiments is the trait-based
261 biodiversity experiment, where the design of the synthetic communities considers not only
262 species' taxonomic identity but also their functional traits. The communities' composition is
263 then defined based on the species traits, aiming for a range of variation in trait composition
264 and diversity (Dias et al., 2013). Scherer-Lorenzen et al. (2007) assembled synthetic forest
265 communities based on traits linked to resource acquisition. Ebeling et al. (2014) did a similar
266 experiment in grassland, which allowed for the isolation of the effect of trait composition

267 and diversity on soil processes (Steinauer et al., 2017) and stability (Fischer et al., 2016).
268 Other studies have adopted similar approaches, such as Galland et al. (2019) and Karimi et
269 al. (2022), who sowed species mixtures presenting independent levels of functional diversity
270 and phylogenetic diversity, and Pichon et al. (2022) who sowed mixtures with different
271 species richness and characterised by slow- or fast-growing species, combined with
272 treatments of N enrichment and pathogen removal. However, studies such as Fukami et al.
273 (2005) and Veen et al. (2018) may have potential confounding effects of species identities,
274 due to the small number of combinations of species for each mixture with a specified
275 functional and phylogenetic diversity.

276 In synthetic community experiments, the process of community assembly takes place
277 over the further development of the communities by means of newly colonising species and
278 changes in abundance or extinction of target species. However, this development is usually
279 not the focus of the study, and there is an effort to remove non-target species in the
280 community to keep only the target ones. This can be challenging, as maintaining the
281 intended composition may not be possible due to extinctions (see Weisser et al. (2017) and
282 Karimi et al. (2022)). In this respect, Grime (2002) argues that “we should not merely
283 review the traits of the plant species in a plant community in order to predict their effects on
284 ecosystem functioning: we should also use them to understand how they were admitted to
285 the vegetation in the first place”. Thus, synthetic community biodiversity experiments help
286 understanding biodiversity effects on ecosystem functioning, but they are not usually used
287 for exploring intrinsic assembly processes, which are mostly halted by artificially removing
288 non-target species. A side-effect of this is the risk of a confounding effect of the disturbance
289 caused by the removal of non-target species. If some mixtures are less resistant to the
290 colonisation of non-target species, they may also suffer a stronger effect of disturbance.
291 Placing this approach in Figure 1a, its focus is on the **Q->F** link, assuming unrealistically

292 that $TC \sim Q$ does not vary during the experiment. Furthermore, most studies with synthetic
293 communities ignore the “natural” state of community development, as the intended
294 parameters of the communities, not the realised ones, are often used as predictors.

295 Yet, we find a few studies based on synthetic community experiments that have
296 evaluated the effects of mixtures of planted species and their traits on different aspects of
297 community assembly, i.e. implicitly following the feedback paths in the conceptual model of
298 Figure 1. Among them, we mention Roscher et al. (2009, 2013, 2014, 2016), who assessed
299 non-weeded plots of the Jena experiment to explore questions related to colonisation and
300 invasion. Galland et al. (2019) assessed the resistance of resident communities to the natural
301 colonisation by species that were not included in mixtures manipulated with different levels
302 of functional diversity and phylogenetic diversity. However, Galland et al. (2019) did not
303 consider this process from the perspective of the resident communities, i.e. how they
304 changed upon the colonisation of new species and in what direction. Interestingly, Pichon et
305 al. (2022) tested hypotheses about the effects of N enrichment and pathogen removal on
306 sowed species mixtures presenting contrasting growth strategies, with a focus on changes in
307 relative species abundances and intraspecific trait variation. Similarly, Karimi et al. (2022)
308 were interested in the temporal dynamics of the communities composed of sowed species
309 mixtures with varying functional diversity and phylogenetic diversity. Nonetheless, since the
310 experimental approaches adopted by Pichon et al. (2022) and Karimi et al. (2022) included
311 the removal of non-target species, it is not possible to disentangle the effects, on community
312 assembly, of the initial functional composition or diversity from those caused by the removal
313 of non-target species followed by the establishment of new ones before the next weeding.

314

315 *Natural community experiments*

316 In contrast with synthetic community experiments, natural community experiments start
317 from pre-existing natural communities, where the previous assembly processes are
318 preserved. Placed within the context of Figure 1, the species composition (**C**) and resulting
319 traits at the community level (**Q**) can be manipulated by removing or adding target
320 components. This manipulation is achieved through the use of biodiversity removal
321 experiments and biodiversity addition experiments, respectively. **C** and **Q** can also be
322 manipulated by altering either the resources, environmental conditions, disturbances or
323 trophic interactions to which the communities are subject. The goal is to manipulate
324 community feedback on ecosystem functions and environmental factors, which in turn affect
325 community assembly – the feedback paths described in Figure 1.

326 Removal experiments are based on the removal of target species, or functional
327 groups, from natural communities (Díaz et al., 2003). We mention the experiment of Mason
328 et al. (2011), which tested the hypothesis that niche overlap in terms of functional traits
329 among the remaining species would decrease with the removal of the dominant species, with
330 consequences on community assembly. Also, the grassland experiment of Joner et al. (2011)
331 tested the hypothesis that the removal of dominant species of the same functional group
332 (only graminoids or only forbs) would reduce functional redundancy and thus the
333 opportunities for species compensation within functional groups in terms of cover
334 (confirmed) and biomass production (not confirmed).

335 Along the same line, the experiment performed by Herben et al. (2013) evaluated the
336 response of the remaining community to plant functional group removal. This study did not
337 support the hypothesis that replacement groups are necessary to maintain biomass
338 production. However, Herben et al. (2013) found differences between groups composed
339 primarily of grasses and groups composed of dicots, suggesting the importance of species

340 differences in the regeneration niche as one of the key traits in the functioning of
341 communities. A similar approach of removing entire functional groups was adopted by
342 Helsen et al. (2016), which suggested the persistence of priority effects in community
343 assembly, as the target species recovered a few years after removal. Lyu et al. (2017)
344 manipulated natural grassland communities by creating plots with different species richness
345 from one to eight, with different species identity combinations, plus the controls with no
346 removal. The focus was on the original intended composition, not on the community
347 assembly process taking place after the removals. Plant traits, though measured, were not
348 considered in the selection of the species.

349 Removal experiments are important tools applied for the study of plant invasions.
350 They can be used to test for biotic resistance to invasion by manipulating certain
351 components of the resident community (e.g. through functional group removal), which is
352 important for predicting alien species establishment (e.g. Byun et al., 2013; Carr et al., 2019;
353 Park et al., 2022; Puritty et al., 2018). The removal of invasive alien species in local
354 communities has been useful to studying the processes of invasion and community
355 reassembly (e.g. Fried et al., 2019; Guido et al., 2021; Guido & Pillar, 2017).

356 Although removal experiments are an interesting approach for gaining insight into
357 the community assembly process, there are methodological limitations that should be
358 considered. Díaz et al. (2003) suggested that the removal effect might be the result of at least
359 three components: (i) the loss of certain traits, which is the focus of most studies, (ii) the
360 response of the remaining plants, depending on which plants occupy the released resources,
361 and (iii) the disturbance effect itself, which involves non-target changes in resource supply.
362 However, most studies did not consider an appropriate treatment control to disentangle these
363 effects, leading to ambiguous interpretations (Guido & Pillar, 2015). To avoid this, it is
364 necessary to include a removal control to distinguish the effects of biomass removal

365 disturbance from the local species extinction effect. This can be done, for example, by
366 removing an equivalent amount of biomass without altering trait community composition
367 (Guido & Pillar, 2015).

368 Biodiversity addition experiments involve sowing (i.e. seed-addition) or planting
369 species into an established community to evaluate the effect of the added components on the
370 assembly process. In most cases, the established community is naturally assembled, and
371 target components are introduced. However, a few studies combined species addition in
372 synthetic community experiments (e.g. Fargione & Tilman, 2005; Roscher et al., 2009,
373 2014). Addition experiments aim to manipulate propagule supply by adding new species'
374 propagules or seedlings to a local community. The added species may belong to the regional
375 species pool or be alien species (e.g. Bennett & Pärtel, 2017, 2021; Breitschwerdt et al.,
376 2015; Kempel et al., 2013; Oster & Eriksson, 2012). The success and abundance of the
377 added species, as well as the dynamics of the resident community after the addition (e.g. an
378 increase/decrease in diversity, or displacement of certain groups), are affected by the
379 similarities between the resident community and the added species in terms of traits
380 (Houseman & Gross, 2011).

381 If a species can successfully establish in the community only when it is
382 experimentally added, then its absence in the local community may be due to dispersal
383 limitation (Houseman & Gross, 2011; Kandlikar et al., 2022). However, if an added species
384 fails to establish despite overcoming propagule limitation, the species trait values may be
385 unsuitable under the prevailing environmental or biotic filtering. This is related to the trait
386 similarities between the added species and the resident community (Bennett & Pärtel, 2017;
387 Breitschwerdt et al., 2015). To evaluate addition experiments in grassland, Bennet & Pärtel
388 (2017, 2021) analysed the composition of species pools from the regional to the site-specific
389 pool and to the local communities to predict the probabilities of a given species to be part of

390 the site-specific pool and to successfully establish in a local (resident) community. For such
391 modelling, they used as predictors the measured trait dissimilarity of each species from each
392 site-specific species pool (or from each local community), and as responses the
393 presence/absence of the species in the site-specific species pool (or in the local community).
394 This multi-scale model of community assembly was useful for predicting species
395 establishment but performed poorly for predicting biotic interactions (Bennett & Pärtel,
396 2021). However, the traits of the species composition observed in a resident community may
397 modify the local environment in a way that prevents certain species from establishing, even
398 if they belong to the site-specific species pool. Therefore, we suggest that the understanding
399 of the failure or success of added species establishment may benefit from the consideration
400 of the feedback path shown in Figure 1.

401 Community assembly studies often use experiments to manipulate environmental
402 factors and assess the selecting effect of the environment on species based on their traits.
403 These experiments typically involve in situ manipulation of resources (such as nutrient
404 addition, rain manipulation, and shading), conditions (such as using open top chambers to
405 control temperature), and disturbance by trophic interactions (such as grazing) or by fire,
406 mostly in natural communities. Another common approach is the turf/monolith-transplant,
407 which involves relocating the entire community to a new environment (e.g. Debouk et al.,
408 2015).

409 While most of these experiments do not control the processes at the level of
410 metacommunity (i.e. available propagules from the surrounding communities), they offer the
411 advantage of being established in natural communities. By altering environmental factors,
412 researchers can evaluate the effects of these changes, or the cessation of a disturbance
413 regime (e.g. grazing) on community assembly through species re-sorting.

414 Environmental manipulation can also be combined with other approaches, such as
415 removal experiments (as shown in Mason et al., 2011). Such integration can enhance our
416 understanding of succession dynamics from degraded to target conditions in restoration and
417 provide insights for improving community recovery, thus bridging the gap between theory
418 and practice (e.g. Funk et al., 2008; Navarro-Cano et al., 2019; Temperon & Hobbs, 2004).

419 One promising yet rare experimental approach is manipulating environmental
420 heterogeneity. For example, Price et al. (2014) conducted a mesocosm experiment to test the
421 effects of soil fertility and heterogeneity on synthetic grassland communities. Through
422 changes in traits related to resource acquisition and competitive ability, they found that soil
423 heterogeneity significantly affected the community-weighted mean (CWM) and niche
424 overlap of co-occurring species. This indicates that the fine-scale distribution of resources in
425 the soil plays an important role in community assembly. Additionally, as shown in Figure 1,
426 heterogeneity may also be created in the process of community assembly. Furthermore,
427 experimental manipulations of environmental factors could be employed to examine
428 hypotheses regarding potential feedback loops illustrated in Figure 1. For instance, a finely
429 tuned, localized addition of nitrogen fertilizer in an experiment could be employed to
430 obscure or mimic the feedback loop generated by the introduction of legumes into the plant
431 communities. Similarly, introducing experimental changes in light exposure or shading
432 could obscure or replicate a feedback loop created by taller plants influencing the
433 availability of light to shorter plants. A good example of this type of experiment is found in
434 Craine & Orians (2004).

435

436 **Community assembly simulation models: a quest for a synthesis**

437 The simulation of community assembly involves proposing causal links between traits and
438 community assembly to guide the simulation process, which will hopefully generate

439 expected patterns in the simulated data. Model validation can be based on the
440 correspondence between the expected and the observed patterns in simulated data. Model
441 validation can also be based on the correspondence between simulated and
442 experimental/observational data, which is a test for the model assumptions. Despite
443 simplifying assumptions, simulation models allow testing hypotheses about how plant traits,
444 environmental factors, species interactions, and metacommunity-level factors such as
445 dispersal limitation influence community composition and diversity. Here we focus on niche
446 models that consider plant selection (filtering) processes involving at least plant traits in a
447 species pool and environmental factors. We focus our discussion on stochastic individual-
448 based models that predict community assembly. Additionally, we highlight models that
449 address the community assembly processes indicated in the conceptual model of Figure 1.

450 Process-based models, such as the dynamic global vegetation models (DGVMs), can
451 simulate the establishment of individuals in plant communities based on ecophysiological
452 processes that link traits or plant functional traits (PFTs) to environmental conditions (e.g.
453 Blanco et al., 2014; Scheiter & Higgins, 2009; Smith et al., 2001). Regarding the simulation
454 of community assembly, these models implicitly model plant interactions by considering the
455 impact of each individual on the available resources, which, in turn, affects the fitness of
456 other individuals in the community, in line with the conceptual causal model depicted in
457 Figure 1. Such a feedback approach is advantageous compared to the use of Lotka-Volterra
458 competition coefficients, which are difficult to obtain (Scheiter et al., 2013; Shipley et al.,
459 2006) and, more importantly, are not linked to the actual resources the plants are competing
460 for. Furthermore, spatially explicit process-based models (e.g. Blanco et al., 2014) can
461 consider the effects of dispersal limitation at the metacommunity level. However, early
462 DGVMs were limited by the very small number of PFTs and by fixed trait values that were
463 often tuned during the model calibration process (Scheiter et al., 2013). Such poor functional

464 diversity does not allow for the study of realistic community assembly processes. In this
465 regard, Scheiter et al. (2013) proposed a flexible trait-based approach for the development of
466 process-based models with a large number of plant types presenting different trait value
467 combinations, which can be filtered by factors considered in the community assembly
468 process. Similarly, Metcalfe et al. (2020) developed a process-based simulation model of
469 annual plant communities, which generated data with emergent diversity patterns that were
470 not explicitly anticipated in the model design. This approach helps closing the gap between
471 process-based models and community assembly.

472 Some community assembly models do not include explicit processes of plant
473 establishment by adopting a statistical approach for predicting the successful establishment
474 of an individual plant given its traits and the local conditions. Among these, the community
475 assembly via trait selection (CATS) model uses as input a matrix of species by traits, and for
476 each site a vector of expected community-weighted mean (CWM) values for the traits. The
477 expected CWMs can be found empirically by fitting regression models of observed CWM
478 on environmental factors (Keddy & Laughlin, 2021; Strahan et al., 2018). In the CATS
479 model, the predicted vector of relative species abundances for each site is then obtained by
480 solving a system of linear constraint equations that (a) the species proportions add to unity
481 and (b) maximise Shannon entropy in the community, and (c) the predicted CWM is closest
482 to the expected CWM. The CATS model has been extended by incorporating a fourth
483 constraint (d) that the relative species abundances must also satisfy a specified prior that can
484 be set to reflect dispersal limitation (Shipley et al., 2012; Sonnier et al., 2010). At a first
485 glance, the maximum entropy restriction seems arbitrary (Keddy & Laughlin, 2021), but the
486 solution with maximum entropy reveals the most likely species composition among the
487 many solutions that would meet conditions (a) and (c) (Shipley et al., 2006). The CATS
488 model is purely driven by environmental filtering and ignores interactions (and the feedback

489 path we propose in Figure 1). Yet, it has been successfully applied for predicting plant
490 communities across environmental gradients (see Keddy & Laughlin, 2021 for a review),
491 though the predictive power declines with species richness (Laughlin & Laughlin, 2013;
492 Sonnier et al., 2010). Keddy & Laughlin (2021) suggest that the maximum entropy solution
493 also pushes towards increased trait divergence and against dominance by few species, thus
494 predicting a species composition that is closer to natural communities. However, no
495 predictions can be made about which factors would drive such trait divergence.

496 Here we focus our attention on the stochastic, individual-based models described by
497 Botta-Dukát & Czúcz (2016) and Pillar et al. (2021), which also adopt a statistical approach
498 for linking traits to environmental factors. These simulation models, with some limitations,
499 can help deciphering the links between plant traits and community assembly according to the
500 conceptual model of Figure 1. Botta-Dukát & Czúcz (2016) used a simple simulation model
501 to test the ability of functional diversity indices to detect trait convergence/divergence
502 patterns generated by community assembly. The model input includes a species pool
503 described by traits related to resource use that have a specified correlation structure; a set of
504 sites along an environmental gradient defining optimal (expected) trait values;
505 environmental filtering parameters that specify the tolerance of each species, given the
506 deviation between its trait values and the expected optimal site-specific trait values; and
507 competition parameters for each trait and pair of species. The trait deviations from the
508 optimal and the competitive pressure that are given by trait differences between species
509 determine the individual survival probability at each site. Therefore, the model of Botta-
510 Dukát & Czúcz (2016) adopts an a priori approach to species interactions, rather than the
511 feedback loop through environmental factors we present in Figure 1. Dispersal limitation is
512 not considered.

513 In line with the conceptual model shown in Figure 1, Pillar et al. (2021) (see also
514 Pillar 2023) described a spatially explicit metacommunity simulation model to explore the
515 effects of the trait correlation structure in the species pool, the number of traits, and the
516 strength of interaction effects of environmental factors on community assembly. The input
517 for the metacommunity simulation model includes a species pool, and the corresponding
518 traits and propagule dispersal parameters. The sites are mapped on the geographical space
519 and described by environmental factors, which may be spatially nested. Seeds arrive at a site
520 with a dispersal probability predicted by the distance from source sites with the same species
521 in the metacommunity, which is applied to a dispersal function with parameters set for each
522 species.

523 The niche basis of the simulation model described in Pillar et al. (2021) (see also
524 Pillar, 2023) lies in a specified pool of species with observed trait values, and a trait space
525 with sites described by expected traits values. The expected trait values are based on
526 imputed or empirically determined linear function parameters (slopes) linking each trait to
527 the one or more environmental factors taken independently or interacting. In this trait space,
528 once propagules arrive, the probability that individuals of a species will recruit and survive
529 at the site will depend on (i) how critical is the trait for the species' fitness, and (ii) how
530 close the species is, in trait space, to the expected optimal trait value at the site. Survival also
531 includes a species-specific density-dependent mortality probability. The process of
532 colonisation and death of individuals is repeated many times and for all sites until saturation.
533 In this process, the feedback of the communities on the environmental factors can be
534 considered (Pillar, 2023). This model has been successfully used to assess the effect of
535 hidden factors in the perception of relevant traits in community assembly (Pillar et al.,
536 2021). Additionally, by simulating spatially nested, including feedback-driven factors, trait
537 divergence was generated at the beta or alpha dimensions, depending on the scale of the

538 community units, which mimics observed patterns in nature (Pillar, 2023). This provides
539 empirical evidence that patterns appearing as plant-to-plant interactions actually emerge
540 through environmental filtering driven by the feedback path proposed in our conceptual
541 model (Figure 1).

542 In the context of community assembly modelling, the species pool can either be a
543 real one (e.g. Metcalfe et al., 2020), where trait patterns may reflect well-known trade-offs
544 that limit the realized trait space (e.g. as described by Díaz et al., 2016), or it can be
545 simulated based on a known trait correlation structure to represent these trade-offs (e.g. as in
546 Dantas de Paula et al., 2021). The role of these trade-offs in community assembly can be
547 assessed in simulation models by examining different levels of trait collinearity and
548 strengths of the filtering factors. As demonstrated in Pillar et al. (2021), one trait may be
549 more limiting than another depending on the strength of the factor effects, but this
550 suppression effect is only present if these traits are independent each other at the individual
551 level, i.e., are not bound by the same trade-off.

552

553 **Conclusions**

554 Our synthesis of concepts and methodological approaches indicates that plant community
555 experiments and simulation models are useful tools for studying community assembly and
556 understanding the underlying processes that generate observed patterns. However, we have
557 identified limitations that hinder a deeper understanding of the links between environmental
558 factors, plant traits, species assembly, and ecosystem functions. We highlight three aspects
559 that, if explored more deeply, could open new avenues in community assembly studies: i)
560 coordination between functional traits, so that environmental filtering selects individuals
561 from a species pool based on non-independent traits bounded by trade-offs; ii) feedback
562 from the existing community, which can modify environmental factors during the assembly

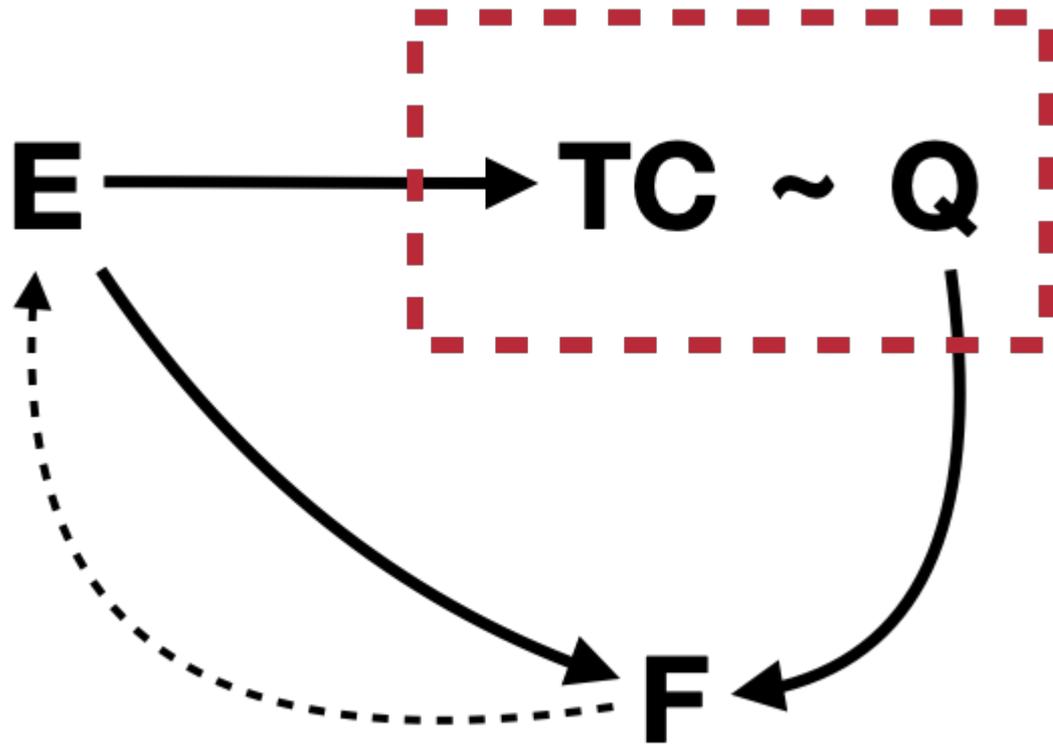
563 process and generate iii) nested environmental variation, which may exhibit distinct spatial
564 and temporal structured at different scales, further producing fine-scale heterogeneity and
565 trait divergence. To address these gaps and guide future studies, we advocate a conceptual
566 model that explicitly indicates the causal links connecting community assembly processes
567 (see Figure 1).

568

569

570

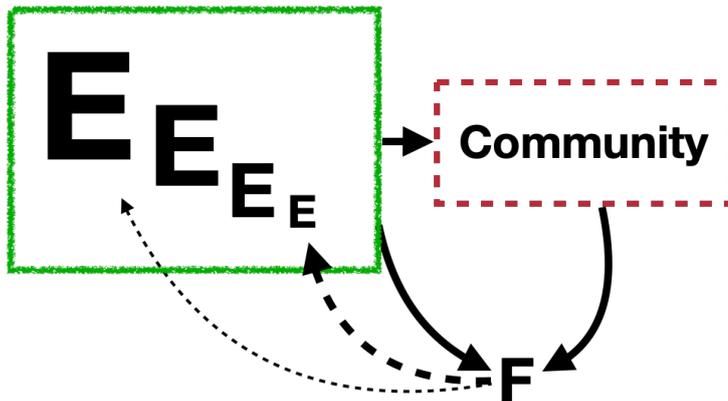
(a)



571

572

(b)



573

574 FIGURE 1. A conceptual causal model that explains community assembly and ecosystem

575 functioning. In (a), each node in the model is represented by a matrix. The communities (C)

576 defined at a given spatial/temporal scale consist of individuals, belonging to a species pool,

577 that have been filtered by one or more environmental factors (E) mediated by one or more

578 traits of the individuals (T). The set of communities in C forms a metacommunity where

579 propagule dispersal occurs. Since the traits and the species cannot be separated, matrices T

580 and **C** are represented as a block matrix **TC** of species by traits and by communities. **Q**
581 expresses traits integrated at the community level (e.g. community-weighted means,
582 functional diversity, fuzzy-weighted species composition), and is computed using the
583 species abundances in **C** and their corresponding traits in **T**, which are represented within
584 the box. The plant communities in **Q** may provide feedback on **E** via ecosystem processes or
585 functions (**F**), indicated by the dashed arrow. As highlighted in (b), the feedback effect
586 depends on the scale of the spatial/temporal variation of the environmental factors in **E**, with
587 stronger effects at finer grains and decreasing effects in coarse grains. Environmental factors
588 in **E** are resources (e.g. light, nutrients, water), conditions (e.g. temperature), disturbances
589 (e.g. herbivory, fire), which are affected by ecosystem functions or processes defined by **F**
590 (e.g. productivity, decomposition, enhanced palatability of flammability). It is important to
591 note that plant-plant interactions are mediated by such feedback, as changes in **E** can
592 influence the community assembly process. Additionally, ecosystem functions (**F**) can be
593 affected directly by environmental factors in (**E**).

594

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611

612 **Authors’ contributions**

613 V.D.P. proposed and designed the study in collaboration with all other authors. All
614 authors (V.D.P., A.G., D.H., F.M.F.) contributed equally to the literature search, screening,
615 and idea development presented in this paper. V.D.P drafted the initial version, and all
616 authors made significant contributions to the writing process and approved the final version
617 for publication.

618

619 **Data Availability Statement**

620 No data was analysed.

621

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