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Towards a new generation of trait-flexible vegetation models

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Fabio Berzaghi^{1,*}, Friedrich J. Bohn^{4,5}, Koen Kramer⁶, Sylvie Oddou-Muratorio⁷,
Christopher P.O. Reyer⁸, Santiago Sabaté^{9,10}, Tanja G.M. Sanders¹¹, Ian J. Wright², Florian
Hartig¹²

9

¹ Laboratory for Sciences of Climate and Environment (LSCE) - UMR CEA/CNRS/UVSQ, Gif-sur-Yvette, France

11

² Department of Biological Sciences, Macquarie University, New South Wales, Australia

12

³ Dipartimento per la Innovazione nei sistemi Biologici, Agroalimentari e Forestali, University of Tuscia, Viterbo, Italy.

14

⁴ Karlsruhe Institute of Technology, Institute for Meteorology and Climate Research, Atmospheric Environmental Research (IMK-IFU), Kreuzeckbahnstrasse 19, Garmisch-Partenkirchen 82467, Germany

17

⁵ Helmholtz Centre for Environmental Research - UFZ, Permoserstraße 15, Leipzig 04318, Germany.

19

⁶ Wageningen University and Research, Droevendaalse steeg 4, 6700AA Wageningen, The Netherlands

21

⁷ URFM, INRA, 84000 Avignon, France

22

⁸ Potsdam Institute for Climate Impact Research (PIK), Member of the Leibniz Association, P.O. Box 60 12 03, D-14412 Potsdam, Germany

24

⁹ Department of Evolutionary Biology, Ecology and Environmental Sciences. University of Barcelona (UB), Barcelona 08028, Spain.

26

¹⁰ CREAM (Center for Ecological Research and Forestry Applications), Cerdanyola del Vallès 08193, Spain.

28

¹¹ Thuenen Institut of Forest Ecosystems, Alfred-Moeller-Str. 1, Haus 41/42, 16225 Eberswalde, Germany

30

¹² Theoretical Ecology, Faculty of Biology and Preclinical Medicine, University of Regensburg, Universitätsstraße 3, 93053, Regensburg, Germany

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* Corresponding author: Berzaghi, F. (fabio.berzaghi@lsce.ipsl.fr)

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1

37 **Abstract**

38 Plant trait variability, emerging from eco-evolutionary dynamics that range from alleles to
39 macroecological scales, is one of the most elusive but possibly most consequential aspects
40 of biodiversity. Plasticity, epigenetics, and genetic diversity are major determinants of how
41 plants will respond to climate change, yet these processes are rarely represented in current
42 vegetation models. We provide an overview of the challenges associated with understanding
43 the causes and consequences of plant trait variability, and review current developments to
44 include plasticity and evolutionary mechanisms in vegetation models. We also present a
45 roadmap of research priorities to develop a next generation of vegetation models with flexible
46 traits. Including trait variability in vegetation models is necessary to better represent
47 biosphere responses to global change.

48

A short review of trait variability and its consequences for ecosystems

Plants are the main primary producers in the terrestrial biosphere and have major impacts on global biogeochemical cycles and climate (e.g. [1–3]). Plant ecosystems also maintain an astonishing amount of biological diversity and provide a multitude of services (e.g. [4]). They do so, at each location, through a mix of growth forms and physiological functions that is known as **phenotypic** or **trait variability** (Glossary) [5].

The study of trait variability has a long tradition in ecology. After all, selection on heritable **intraspecific trait variability** was the key idea for Darwin’s theory of evolution. In the 1970-80s, the emerging field of comparative ecology emphasized plant **ecological strategies** and analyses of **interspecific trait variability** and **trade-offs** between species [6–8]. Energetic organization of trait data during the last 20 years [9–14] has enabled broad-scale quantification of key traits and trade-offs across space and time, and across taxonomic and functional groups [15,16]. A fundamental insight is that certain combinations of traits jointly characterize plant life-histories and their ecological strategies [17,18]. Increasingly, the focus is shifting to the effects of both intraspecific variability [19–21] and **rapid evolution** [22] in the context of species and ecosystem **adaptation** to climate change.

Observed within- and between-species variability in **plant functional traits** can result from **plastic** responses to biotic and abiotic conditions, and from heritable genetic or **epigenetic** differences (e.g., [20,23–26] and Figure 1). Evidence for within-species variability from **provenance trials, common garden experiments** [27–30], and observations of rapid evolution [30–32] demonstrate that intraspecific phenotypic differences are often heritable and adaptive. This suggests that traits can evolve at ecological time-scales and that the concept of a species as a functionally static unit is an approximation at best.

76 A challenge for understanding the consequences of trait variability is that we seldom know if
77 measured trait differences arise from **standing diversity**, plasticity, or both. There is hope
78 that population genomics and next generation sequencing [33] could establish **genotype-to-**
79 **phenotype** maps [34] and make it easier to decipher the genetic and plastic components of
80 observed patterns of variation. Currently, however, most data only document existing
81 patterns of trait variability, with no guarantee that this variability will persist over time, e.g.
82 under changing climates.

83
84 Another key challenge is understanding the consequences of trait variability. Trait variability
85 is often assumed critical for competition, **fitness**, adaptation and resilience in plant
86 communities, particularly during rapid climatic change [35,36] – e.g. because trait variability
87 likely increases **evolvability**, and buffers species and ecosystems against environmental
88 fluctuations [20,23,26]. However, theoretical models show that trait variability is not always
89 beneficial [37] and empirical studies often fail to demonstrate the expected links to fitness
90 [38]. The consequences of trait variability thus seem more complex and context-dependent
91 than commonly assumed. If true, an obvious and possibly the only route forward is to explore
92 these questions with models that describe traits in a more quantitative and mechanistic way.

93 **Modelling the consequences and evolution of plant trait variability: the crucial role of** 94 **vegetation models**

95 Detailed **mechanistic models** already exist that describe communities of plant species in
96 terms of their underlying biogeochemical and ecological processes (**Dynamic vegetation**
97 **models** = DVM see Box 1, see also [39–41]). These models have been instrumental for
98 understanding global biome distributions and for creating dynamic projections of plant
99 ecosystems under global environmental change [41–44].

101 An increasingly noted shortcoming of these models, however, is their limited representation
102 of phenotypic variability (e.g. [45,46]). Most DVMs operate with a fixed phenotype per species
103 or represent the world's species via a small number of **Plant Functional Types** (PFTs). This
104 simplistic representation largely neglects the complex patterns of trait variability within and
105 between species. Such patterns occur at all ecological scales [18,20] and across ontogenetic
106 stages. Fixed phenotypes do not account for changes in traits that may occur through plastic
107 and **eco-evolutionary dynamics** [15,27]. Thus, most current models are ill-equipped to
108 explore the consequences of trait variability for ecosystems dynamics or biosphere-
109 atmosphere interactions [1,3]. The lack of processes responsible for trait variability (Fig. 1)
110 increases the uncertainties in the predictions provided by DVMs.

111
112 In response to these issues, there have been several attempts to establish next-generation
113 vegetation models, which here we summarize as *plastic*, *prescribed*, and *eco-evolutionary*
114 models of trait variability (Box 2 and Supplementary material)

115
116 Plastic models of intraspecific variability emphasize functional plasticity and trade-offs within
117 plants. In such models, plants can, for example, exhibit morphological and/or physiological
118 changes (e.g. in leaf characteristics or allocation strategies) in response to environmental
119 conditions [47–49]. However, these models are limited by the lack of heritable variability. For
120 example, they assume that all individuals of a species in a given environment express the
121 same phenotype and that the adaptation to changes in environmental drivers is
122 instantaneous.

123
124 Models with prescribed intraspecific variability incorporate non-plastic trait variability into the
125 modeled processes, but do not explain its origin. These models may help to understand the
126 role of trait variability in structuring plant communities; however, their ability to correctly
127 predict the re-assembly of communities after disturbance is being challenged. Mounting

128 evidence suggests that evolutionary processes can happen on ecological time scales, in
129 particular under strong selection [31,32].

130
131 Eco-evolutionary models address this shortcoming by simulating intraspecific variability as
132 the result of ecological and evolutionary processes (in particular **natural selection**, e.g. [50]).
133 Simple eco-evolutionary models draw new individuals from a given trait distribution and let
134 selection act upon them. More sophisticated (and computationally costly) models draw new
135 individuals from the current community following genetic inheritance rules, and via explicit
136 models of genotype-phenotype relationships (see Box 3). By representing genetic processes,
137 rapid evolution in response to climate change can be more reliably modeled.

138 **Research priorities for next-generation trait-flexible models**

139 Despite encouraging progress, there is clear potential for a more robust representation of
140 intraspecific trait variability and its genetic origin in DVMs. Below we outline what we see as
141 research priorities and key challenges for creating a new generation of *trait-flexible* models
142 that, in principle, would fully account for and predict observed variability and plasticity in plant
143 traits and functions.

144 Understanding plasticity, acclimation, and lagged effects

145 Current DVMs may include limited plasticity at the process-level, e.g. in allocation,
146 phenology, photosynthesis and autotrophic respiration [51]. In working towards a more
147 complete treatment of physiological and morphological plasticity a first challenge is to create
148 appropriate empirical models. In particular, plasticity as a within-generation response to
149 environmental variability needs to be disentangled from across-generation responses
150 achieved either by epigenetic or genetic inheritance. Representing within-generation lagged
151 effects is especially challenging, e.g. the effect of past stress (drought, starvation, etc.) on
152 plant traits and future mortality rates. Lagged effects are observed in empirical studies but not

153 well understood and thus rarely modeled [52]. A first step would be to incorporate some of
154 these lagged effects through dynamic functions that relate disturbance events to
155 physiological responses [52–54].

156 Closing the demographic loop

157 For modelling eco-evolutionary dynamics, we also have to determine the fitness of any given
158 phenotype or genotype. Many existing vegetation models are not particularly well-suited for
159 this task [39] also due to a limited description of mortality and recruitment processes [39,55].
160 arguably because the primary focus of past model development has been on growth
161 (individual-tree and stand-scale models), productivity, and carbon cycling (global models).
162 Many widely-used models simulate recruitment via a constant seed rain, making reproduction
163 independent from the performance of traits or species in the standing community (but see
164 [56]). This is clearly inadequate to simulate evolution.

165
166 Moreover, a lot is known about variability in seed production and seed size and the
167 implications for plant demography (e.g., [8,17,57]), yet relatively little use is made of this
168 knowledge (but see [58,59]). Other entry points to simulate plasticity across life stages are
169 changes in allocation to plant defenses [60] and ontogenetic shifts in leaf traits [61].
170 Ultimately, without realistic descriptions of all three demographic processes – reproduction,
171 growth, and mortality (i.e., “closing the demographic loop”) – we cannot generate realistic
172 predictions about how selection influences the distribution of observed phenotypes at local
173 scales, especially in variable environments.

174 Space and dispersal

175 Evolutionary models often assume well-mixed populations but we know that genetic and
176 phenotypic variation in plants show strong spatial structure. Few models incorporate key
177 mechanisms for predicting shifts in species’ range [62]. Further, spatial processes are needed

178 to study plant responses when climatic and non-climatic agents of selection are decoupled
179 [63]. When trying to represent these structures and processes in DVMs, we face the problem
180 that most DVMs are not spatially explicit. Typically, they produce maps based on
181 independently-simulated communities (e.g., from 25 x 25m up to several hectares), which are
182 then averaged to create large-scale maps. In principle, the knowledge and data for a better
183 representation of spatial processes is often available. Yet, it can be prohibitively complex to
184 track pollen and seeds or to follow the germination and growth of a large number of offspring.
185 However, recent examples show some solutions to implement spatial processes such as
186 dispersal and pollination at various geographical scales (i.e., local to continental) [64–67].

187 Trade-offs

188 Any eco-evolutionary model will have to define appropriate trade-offs to constrain species
189 properties to eco-physiologically realistic values, and to avoid unbeatable (and unrealistic)
190 “superspecies” (or supergenotypes) emerging and taking over. A pragmatic approach [46,56]
191 is to represent key trade-offs using observed trait correlations (e.g., the leaf economics
192 spectrum or LES: [13]). Another option is to incorporate theory for specific trade-offs, e.g.
193 concerning seed and seedling competition vs. colonization abilities [57], water and nutrient
194 use in photosynthesis [68], or biomass allocation to different plant parts (e.g., [69]). A further
195 approach is to better define trait-climate relationships using combined trait and flux data,
196 allowing one to develop dynamic **acclimation** functions [70]. Still, some fundamental
197 questions remain; most notably the extent to which trade-offs among traits are general within
198 species as well as across taxonomic groups, and independent of other plant parameters [19].
199 Increasingly detailed quantification of geographic trait variation (e.g. [71]) will facilitate a
200 better representation of local adaptation [72] and overcome some of the limitations described
201 in Box 2.

202 Epigenetic inheritance

203 Current DVMs do not consider that plastic responses can be propagated across generations
204 via epigenetic mechanisms [73]. These effects are probably more significant and have a
205 shorter response time than genetically inherited effects [22,63,73] . The main challenge of
206 implementing epigenetic processes is our lack of precise knowledge about them: there is
207 abundant empirical evidence for epigenetic regulation in non-natural environments, but we
208 lack data to make predictive models for natural situations. Limited knowledge of epigenetic
209 mechanisms makes their implementation a low priority, until high-throughput sequencing
210 provides sufficiently abundant quantitative data linking them to evolutionary processes [73].

211 Genetic architecture and the genotype-phenotype link

212 Many eco-evolutionary models of trait evolution assume that mutations act directly on traits
213 [56,74,75] . In reality, mutations act on genes and thus only indirectly on traits via genotype-
214 phenotype links. Accounting for recombination, **genetic architecture** of traits, and any
215 deviation from random mating can lead to important differences in the rate of adaptation
216 simulated by gene-based-evolution models compared to trait-based-evolution models. In
217 particular, recombination enables considerably faster creation of new phenotypes from an
218 existing gene pool than does mutation. The ways in which genetic architecture drives trait
219 evolution is still debated. However, enough knowledge is available to couple basic
220 evolutionary models with explicit genetic structure to DVMs and other ecological models
221 [50,67,76] (Box 3). Presumably, accounting for the rate of adaptation of traits is most
222 important: 1) for short-lived plants; 2) in ecosystems with frequent generational changes due
223 to high disturbance rates or forest management; and 3) for tree populations at the edge of
224 their distributions [77]. Yet, a key problem in these models is predicting phenotypes from
225 genotypes and environment; this requires data and knowledge still unavailable for many tree
226 species and traits of interest, and is often population-specific. Another challenge is scaling up
227 these processes to large spatial and time scales to understand adaptation processes such as

228 gene flow and reproductive isolation across species' ranges (see also next section: *Efficient*
229 *scaling*)

230 Efficient scaling across several space or time scales

231 Implementing many of the processes described above invokes an old, but still unresolved
232 issue in theoretical ecology: the problem of upscaling the effects of local or short-term
233 processes (dispersal, microevolution) to large temporal and spatial scales [78]. Aside from the
234 inevitable technical considerations, it is important to consider that patterns of trait variability
235 might not hold across geographic and ecological scales ([19,79]). These findings have
236 implications for DVMs because global trait parameters, even if flexible, might add uncertainty
237 when upscaling local ecosystem properties. More fundamental research is needed on the
238 subject. For now, a practical and partial solution would be to use georeferenced trait data
239 when determining tradeoffs and, when available, local trait data (see Tradeoff section).

240 Model-Data integration

241 Finally, a crucial point for the utility of new model structures is their ability to use and
242 integrate existing data. General approaches for model-data integration have been discussed
243 elsewhere (e.g. [40]). The core of this discussion is that data can enter the model at different
244 levels - as inputs, as drivers, as prior information on **model parameters** or on model outputs,
245 the latter requiring inverse modelling to back-propagate output data to infer model
246 parameters or states. These general insights remain valid for the problem we tackle here, but
247 in practice the question will be if sufficient data are available to constrain DVMs and where in
248 these models are the data best placed to constrain key processes.

249

250 The most obvious data type to consider is trait data. For example, the TRY database
251 (www.try-db.org) currently holds ca. 12 million trait records from 280000 species. For further
252 trait databases see Schneider et al. 2019 [14], *eFlower* (<http://eflower.myspecies.info/>), and

253 *China Plant Trait Database* [80]. Not all trait records map 1:1 to model parameters, but these
254 databases are clearly a key data source for creating priors on model parameters and trade-
255 offs, or to obtain data on outputs for those model types in which trait distributions are
256 emergent. There are many other, more specialized data types that are also interesting. Data
257 from **provenance** experiments, for example, contain information about heritable intraspecific
258 differences, in particular regarding leaf properties (morphology and chemistry), bud burst,
259 phenology, photosynthetic and hydraulic physiology and other quantitative traits [81]. Both
260 historical and current data from provenance trials are being used for quantitative genetic
261 studies, but these data are underutilized for parameterizing models.

262 **Concluding remarks and future directions**

263 Most current DVMs describe species or PFTs by a single set of properties that is static in
264 space and time. Empirical data show this is a crude approximation at best, as genetic and
265 phenotypic diversity allow for plastic responses and long-term adaptations to environmental
266 conditions, also via epigenetic inheritance.

267
268 We suggest that a new generation of trait-flexible vegetation models is needed which
269 embraces the variability and adaptability of functional traits in vegetated ecosystems. An
270 ideal model would explicitly account for phenotypic plasticity as well as genetic and
271 epigenetic mechanisms in a spatio-temporal context. Such a model would allow researchers
272 not only to quantify the degree to which trait variability buffers diversity and ecosystem
273 functioning against climate change, but also to improve our mechanistic understanding of the
274 processes contributing to trait diversity, and thus biodiversity. Trait-flexible models could help
275 separate the heritable components of trait variability from plastic components, across-
276 generational trait variability, and the influence of biotic and abiotic factors. Inclusion of a wider
277 range of flexible traits could further help determining the individual contribution of certain
278 traits to plant fitness.

279

280 One risk of adding additional processes to existing models is that they become increasingly
281 complex, difficult to parameterize, and thus be prone to higher uncertainty [82] [83].

282 Countering that is the potential to increasingly rely on optimality-based approaches in model
283 components. In principle, these approaches should allow models to be at least as powerful
284 as standard approaches but would require fewer inputs (e.g. [68,69,84,85]). In practice,
285 compromises will have to be made between accuracy, realism, and generality [83] (see also
286 Outstanding Questions), but we believe that immediate progress is possible in the areas that
287 we have highlighted in this Opinion piece.

288

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290

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- 635

636 **Figure legends**

637 **Figure 1 - Conceptual figure of trait variability.** Rectangles represent states (genotype or
638 phenotype) and ovals represent processes driving variability. The top half represents
639 individual-level processes determining trait variability within a lifetime. The bottom half
640 represents community- or population-level processes that influence variability between
641 generations. Dashed arrows represent how genetic and plasticity can trigger changes in
642 phenotype within a lifetime of an individual.

643

644 **Figure I Box 3: Physio-demo-genetic (PDG) models: outline and an application. A)**

645 Conceptual framework of PDG models. PDG models couple: (i) a biophysical module to simulate
646 carbon and water fluxes at the tree level using climate observations; (ii) a forest dynamics module
647 to calculate demographic rates for adult trees (growth, mortality, and reproduction) based on
648 carbohydrate reserves, and to simulate ecological processes across the life cycle; and, (iii) a
649 quantitative genetics module relating genotype to the phenotype of one or more functional traits.

650 B) (left) Latitudinal cline of chilling (i.e. low temperature) requirements (Sc^*) and (right)
651 dependency of forcing (i.e. high temperature) requirements (Sf^*) on average temperature after 400
652 years of simulation. Initially, both Sc^* and Sf^* start with a single value at all locations. (reproduced
653 from [76]).

654

657 **Box 1 Dynamic vegetation models – an overview**

658 The overarching goal of dynamic vegetation models is to describe the functioning of vegetated
659 ecosystems by modelling the key processes that drive vegetation dynamics: primary production,
660 competition, water and nutrient cycles (Figure I, see also [40,41]). These processes are simulated in
661 response to abiotic and biotic drivers.

662
663 Models differ in their emphasis on different processes and scales. Some models concentrate on
664 shorter time scales, with a focus on productivity and allocation as well as short-term feedbacks such
665 as water and nutrients. Other models also consider succession and changes in community structure
666 as a result of mortality, competition, and disturbances [55]. Global models are often used to study
667 general patterns in vegetation composition or can be coupled to climate models to examine
668 vegetation-atmosphere feedbacks [1]. At the local end of the spectrum, we find forest models focusing
669 on ecophysiology, adaptation, and forest management [41,55]. Despite these differences in the scale
670 of interest, most models create their predictions by simulating a local community or ecosystem, which
671 is then upscaled to the desired resolution. As a result, most global models can and are also used to
672 model stand-scale dynamics. Still, most models do not explicitly consider spatial feedbacks and
673 dynamics. An exception are so-called landscape models, which are spatially explicit and can
674 reproduce spatial processes such as fire or insect outbreaks [64].

675
676 Functional diversity is typically implemented by assigning different model parameters for each species
677 or **plant functional type** (PFT). The main aim of the PFT concept is to reduce the, often large,
678 taxonomic diversity to a small number of manageable vegetation types, such as deciduous temperate
679 forests, evergreen boreal forests, continental grasslands, etc [86]. An alternative to this species or
680 type-centered paradigm are trait-based models. These drop the species concept in favor of describing
681 individual plants by a set of traits, emerging from general correlations and dependencies such as the
682 leaf and stem economics spectra (e.g., [46,56,87]).

BOX 2: State-of-the-art of models including intraspecific trait variability

Within the ecological modelling literature, we identified three main approaches for including plastic

trait variability. See supplementary material for a complete list of models and methods.

1. Plastic models of intraspecific variability	
General approach	Plasticity is described through equations of the physiological response to a change in environmental conditions. For example, acclimation of photosynthesis is described by an equation describing the response of maximum photosynthetic rate in relation to leaf nitrogen concentration [48].
Advantages	<ul style="list-style-type: none"> - Allows plant responses to be simulated in different environments and ecosystems, as equations are used to describe the general functioning of plant ecophysiology.
Limitations	<ul style="list-style-type: none"> - Difficult to determine generalizable functions applicable to different plants and conditions. - By describing only one plastic response, related counter-responses might be missing because physiological processes are interrelated.
Examples	PnET-CN [49]; IBM-Esther [47]; ORCHIDEE-NP [48]
2. Models based on prescribed trait distributions	
General approach	Well-established trade-offs (e.g., leaf- and wood-economic spectra) are used to assign and constrain trait values corresponding to model parameters. Trait values are sampled from empirical trait distribution generated from plant trait databases (e.g. [14]) or local measurements. Different trait values are assigned to individual plants.

Advantages	<ul style="list-style-type: none"> - No major modification of model structure necessary. - Allows studying the effects of selection and trait variability. - Many data are available for traits across the leaf and stem-economic spectrum.
Limitations	<ul style="list-style-type: none"> - Ongoing adaptation to changing local environmental conditions is not simulated, which implies no trait evolution. - No trade-offs or constraints, potentially leading to unrealistic combinations of traits - A more detailed description of intraspecific / interspecific variability is impossible due to the lack of a mechanistic underpinning corresponding to the real ecological processes. This limitation makes predictions from such a model somewhat questionable. - Not bound to any particular spatial scale and can potentially represent a large number of combinations of traits but it is not clear if these combinations represent actual species.
Examples	ArcVeg [88]; Ecotone [89]; CABLE [90].
3. Eco-evolutionary models (species or trait-based)	
General approach	<p>Concepts from quantitative genetics are used to simulate adaptive evolution; trait values are determined by the genetic contribution of one to several loci associated with a particular trait. Trait distributions emerge from some or all of the following processes: heritability, mutation, gene flow. A detailed example is presented in Box 3.</p>
Advantages	<p><i>Trait- and species-based models:</i></p> <ul style="list-style-type: none"> - Simulates the rate of adaptation to changes of environmental drivers from first principles - Allows simulating adaptive responses of plastic and non-plastic traits - The equations controlling the quantitative genetics are first order algebraic

	<p>(easily computable).</p> <p><i>Trait-based models only:</i></p> <ul style="list-style-type: none"> - Avoids technical and data limitations associated with working at the species level. While species-level trait data are often difficult to obtain, community-level relationships are relatively well described. - Inter- and intraspecific variability are both covered - By applying selection on the initial trait space, functional diversity patterns at the macro scale emerge, e.g. observed clines in traits across environmental gradients.
<p>Limitations</p>	<ul style="list-style-type: none"> - Requires insight of genetic architecture of traits (distribution of allelic effects in the population, number of alleles and loci determining the trait, interactions between alleles and/or loci); Until now, only simple genotype-phenotype relationships are included (e.g., no epigenetics or gene activation) - Requires knowledge on pollen and seed dispersal as well as mating system - Because of computation costs, only a subset of traits can evolve, and need to be selected. <p><i>Trait-based models only:</i></p> <ul style="list-style-type: none"> - Plants are defined by sets of traits not directly attributable to any species, difficult to connect to species-specific data for validation and initialization purposes.
<p>Examples</p>	<p>aDGVM [56]; FORGEM [76]; PDG [67]; Jedi-DGVM [87]</p>

691
692

693 **Box 3 - Integrating physiology, demography and genetics in forest trees**
694 **with Physio-demo-genetic models**

695
696
697 Physio-demo-genetic (PDG) models integrate physiological, demographic, and evolutionary
698 processes. PDGs have been developed to better understand the interplay among plasticity
699 and genetic adaptation and the effects of both processes on tree population dynamics under
700 global change [67,76,91]. The advantage of PDG models is their ability to account for the
701 variability in functional traits due to both standing genetic variation and evolutionary change
702 in response to changing local environmental conditions.

703
704 In such models, trait values are modeled following a classical quantitative genetic model
705 (Figure IA) and are thus inherited by a tree's offspring. As reproduction, growth, and survival
706 of individual trees depend on their particular trait constellation, the interaction between the
707 biophysical- and the demographic model results in a trait-fitness relationship for the particular
708 environmental conditions. PDGs have been applied to examine the adaptation of the timing
709 of budburst (TBB) along altitudinal or latitudinal gradients of *Fagus sylvatica* (the European
710 beech). Simulations show that few generations were sufficient to develop non-monotonic
711 genetic differentiation in the TBB along the local climatic gradient (Figure IB).

713 Glossary Box

714 **Acclimatization/acclimation:** plastic response of an individual to environmental change
715 within its lifetime, often important on short timescales (e.g., months/seasons). We consider
716 the two terms as synonyms in the broadest sense, without distinction between natural or
717 controlled environments.

718 **Adaptation:** genetic response to selection resulting in genetic and phenotypic changes
719 between generations. Natural selection relies on differential fitness of individuals with different
720 phenotypes.

721 **Common garden experiment / provenance trial:** a classic experimental design to test for
722 heritable differences between populations/provenances, consisting of planting seedlings or
723 seeds of the same family or species, but from different geographical origin or context, in the
724 same location.

725 **Dynamic vegetation models (DVM, related: land surface models, ecosystem models,**
726 **process-based vegetation models, terrestrial biosphere models, etc.):** a group of
727 models that dynamically simulate the succession, distribution, and structure of natural
728 vegetation, using mostly mechanistic representations of large-scale vegetation processes.

729 **Eco-evolutionary dynamics:** the interplay of ecological and evolutionary processes at the
730 same time scales.

731 **Ecological strategy:** the manner in which a species obtains and uses resources, interacts
732 with other organisms, copes with (and is adapted to) environmental stresses and, in the end,
733 ensures genetic continuity among generations.

734 **Epigenetic:** heritable changes in gene expression that are not caused by mutations of the
735 DNA sequence. The induced phenotypic changes (due for instance to DNA methylation) are
736 potentially reversible, and thus less stable than changes due to DNA sequence mutation.

737 **Evolvability:** the capacity of a biological system for adaptive evolution, and thus the ability to
738 adapt to future environments through evolutionary processes.

739 **Fitness:** expected lifetime reproductive success of a phenotype or genotype, frequently
740 expressed as a result of abiotic and biotic processes.

741 **Genetic architecture:** the number of loci determining a given trait and whether a locus acts
742 on several traits (pleiotropy); the number of allele per loci and their effects on each trait, the
743 degree of linkage between loci, and interactions between loci (epistasis).

744 **Genotype:** a set of genes that determine the phenotype.

745 **Mating system:** determines who mates with whom in a species or population, and includes
746 the self-pollination rate, the rate of pollen immigration and the variances of individual
747 fecundities. The mating system determines the level of genetic drift, the level of inbreeding of
748 the offspring, and, through inbreeding depression, the dynamics of natural regeneration.

749 **Model parameter:** a variable in an equation or algorithm describing (part of) a mechanism or
750 process. In dynamic vegetation models, parameters can define traits or properties of
751 processes included in the model.

752 **Phenotype:** a set of traits

753 **Trait or phenotypic variability:** the variation in phenotypes between species (interspecific)
754 or within and among populations of a species (intraspecific) which can arise from genetic and
755 epigenetic variation, and/or from plastic responses to the environment.

756 **Plant functional traits:** observable characteristics of a plant, including morphological,
757 physiological and phenological characteristics. Traits influence the demographic and
758 reproductive performances or ecological functions of a plant.

759 **Plant Functional Type (PFT):** The PFT concept classically refers to aggregating functionally
760 similar species into a single group (i.e., functional type). In principle, PFTs could also be
761 defined to classify variability within a species.

762 **Plasticity:** the phenomenon of the same genotype producing different phenotypes in
763 response to different environmental stimuli. Plasticity includes acclimatization/acclimation.

764 **Provenance:** variety of a tree species from a well-defined geographic area.

765 **Rapid evolution:** the response to selection within a few generations, e.g. through selection
766 on **standing diversity** and recombination.

767 **(Natural) selection:** the main mechanism of evolution relying on differential survival and/or
768 reproductive success of individuals with different phenotypes. Selection affects the standing
769 diversity.

770 **Standing diversity:** existing genetic diversity within a population.

771 **Trade-off:** trade-offs describe situations where further investment of resources in some trait
772 or process necessitates less investment in another trait or process.

773 **TABLE 1 - Traits with their reported sources of intraspecies variability categorized by class**
774 **or process type, followed by key empirical correlations or constraints and DVMs including**
775 **a trait-flexible implementation ^{a,b,c,d}. The full list of models and details on their**
776 **implementation of flexible traits can be found in the supplementary material.**

Class/ Process	Trait	Reported within- species variation: source not distinguish ed	Evidence for genetic/ epigenetic link to variation	Variation from plasticity	Empirical key correlations or constraints with other traits or processes	Trait variability in current models
Photosynth esis	Maximum photosynthet ic rate, V_{cmax}	[92]	[93]	[93–96]	Positively correlated with leaf nitrogen, phosphorou s and specific leaf area [94,97– 99]	LPJmL-FIT [46], MATEY [85]
	Leaf mass per area (g m^{-2}) or its inverse, specific leaf area (cm^2g^{-1})	[100–103]	[27,104,104– 107]	[95]	Correlated negatively with leaf lifespan, positively with leaf nitrogen, phosphorou s and dark respiration [13,86]. Higher at more arid sites and higher irradiance	LPJmL-fit [46], MATEY [85], LM3- PPA [110]

					sites [108,109]	
	Nitrogen content (g m ⁻²)	[102,103]	[27,107,111]	[71]	Higher at more arid sites, higher irradiance [108,112]	CABLE [90], MATEY [85], TFS [113]
	Stomatal conductance (gw)	[114]	[27,115]	[116,117]	Assimilation rate higher at low soil phosphorus [118]	-
Respiration	Proportional change in respiration per 10 C increase in temperature (Q ₁₀)	[119]	-	[120,121]	Two types of acclimation: 1- adjustments in the Q ₁₀ and 2- changes in the enzymatic capacity of the respiratory system.	PnET-CN [49]
	Number of mitochondria per cell	[122]	-	[123]	Respiration, growth, acclimation at high CO ₂ [123].	-
	Leaf respiration in darkness and in the light	[103]	[27]	[95,120]	Leaf dark respiration varies in relation to site climate [124].	LM3-PPA [110]
Wood/ leaf	Wood density	[125,126]	[104,127,128]	[129]	Connections to growth	LPJmL-FIT [46],

structure					rate and mortality. Higher at more arid sites [125].	aDGVM2 [56], TFS [113]
	Wood and sapwood hydraulic conductivity and anatomy (vessel Lumen area, ring-width). Hydraulic plasticity and P50 ^d	[126,130]	[104,128]	[131–135]	Globally, higher at wetter sites, higher at warmer sites [136,137]	-
	Herbivory defense (plant secondary metabolites, spines) or serotiny (fire resistance)	[138]	[138–141]	[142]	Defense-growth trade-off influences allocation strategies [143].	-
	Bark thickness	[144]	[145,146]	[145]	Thicker in sites with more frequent fire and/or more frequent/severe drought, and hotter sites [144,147,148]	LPX [149]
Phenology	Timing of budburst	[150]	[151–153]	[151]	constraints: duration	Forgem [76], PDG [67]

					growing season vs frost damage [24,153]	
	Timing of leaf fall / growth cessation	[154]	[24]	[155]	constraints: duration growing season vs frost damage [24,153]. Nutrient resorption [154]	-
	Frost hardiness	-	[27]	[156,157]	constraints: duration growing season vs frost damage [156]	Forgem, Forgro [76]
Allocation	Biomass compartments, allometries (leaf: fine root, diameter: height)	[158]	[27,105,111]	[111,159]	Growth, stem and leaf economics spectra	Jedi-DGVM [87], aDGVM2 [56]
	Leaf area / sapwood area ratio	[126]	[104]	[131,132]	Globally, lower at drier sites, and in species with smaller leaves and/or lower specific leaf	-

					area [137]	
Reproductio n	Seed size	[160,161]	[162]	[57,163]	Height, seedling survival, dispersal distance [8,160].	hybrid-DVM [164], aDGMV2 [56]
	Germination rate	[161]	[27,105]	[165]	Recruitment	Esther IBM [47]

777

778

779

^a The source of variability, when determined, is classified as plastic or genetic or both. It also provides a brief description of key correlation with other traits and processes.

780

781

^b A list of examples is presented in which dynamic vegetation models have implemented trait variability for the specified trait.

782

783

^c This table is not an exhaustive list, thus more references in a cell do not necessarily imply a

784

higher number of studies, when possible at least 2 examples per cell were included. Primary

785

literature was preferred over review articles. The full list of models and reviewed articles can be

786

found in the supplementary material.

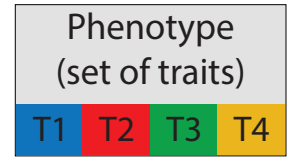
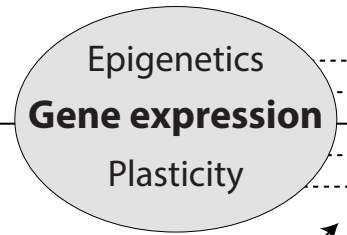
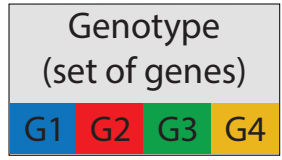
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^d P50: water potential at which 50% of hydraulic conductivity is lost

Individual level

Community / species level

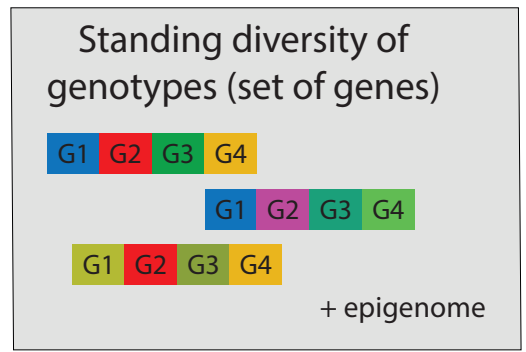
Within-lifetime processes



Biotic and abiotic
Environment

determines

determines



Between-lifetime processes

