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The Latitudinal Diversity Gradient: Novel Understanding through Mechanistic Eco-evolutionary Models

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Abstract: The latitudinal diversity gradient (LDG) is one of the most widely studied patterns in ecology, yet no consensus has been reached about its underlying causes. We argue that the reasons are the verbal nature of existing hypotheses, the failure to mechanistically link interacting ecological and evolutionary processes to the LDG, and the fact that empirical patterns are often consistent with multiple explanations. To address this issue, we synthesize current LDG hypotheses, uncovering their eco-evolutionary mechanisms, hidden assumptions, and commonalities. Furthermore, we propose mechanistic eco-evolutionary modeling and an inferential approach that makes use of geographic, phylogenetic, and trait-based patterns to assess the relative importance of different processes for generating the LDG.

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52 Abstract

The latitudinal diversity gradient (LDG) is one of the most widely studied patterns in ecology, 53 54 yet no consensus has been reached about its underlying causes. We argue that the reasons are the verbal nature of existing hypotheses, the failure to mechanistically link interacting 55 ecological and evolutionary processes to the LDG, and the fact that empirical patterns are 56 57 often consistent with multiple explanations. To address this issue, we synthesize current LDG hypotheses, uncovering their eco-evolutionary mechanisms, hidden assumptions, and 58 commonalities. Furthermore, we propose mechanistic eco-evolutionary modeling and an 59 inferential approach that makes use of geographic, phylogenetic, and trait-based patterns to 60 assess the relative importance of different processes for generating the LDG. 61

62

63 State of the art and calls for novel mechanistic approaches

The increase in species diversity from the poles to the equator, commonly referred to as the 64 latitudinal diversity gradient (LDG), is one of the most pervasive [1, 2] and widely debated 65 biological patterns, with at least 26 listed hypotheses associated with it [3-5]. These 66 67 hypotheses can be classified into three higher-level categories related to latitudinal variation in ecological limits (See Glossary), diversification rates, and time for species accumulation 68 (Table 1). Empirical evidence seems compatible with many of these hypotheses. For example, 69 species richness is correlated with purported proxies for ecological limits such as net primary 70 productivity [6-8], diversification rate can vary latitudinally due to gradients in temperature 71 [9, 10], and diversity is greatest in regions where diversification has occurred over a longer 72 period [11-13]. These and similar studies have improved our understanding of the LDG and 73 macroevolutionary patterns in general, but the diffuse support for different hypotheses reveals 74 a lack of consensus and points to challenges in testing and evaluating these hypotheses. 75

We argue that reconciling the causes of the LDG requires moving beyond verbal chains of 77 78 logic, which are inherently prone to error with respect to how assumptions result in their predicted effect [14], and towards a more formal and mechanistic framework. Verbal 79 hypotheses often contain hidden assumptions that go untested and lack specificity with 80 respect to the mechanistic underpinning of relevant ecological and evolutionary processes. 81 Verbal hypotheses also tend to focus on a single driver to predict just one or a few patterns 82 related to that driver. Consequently, these predictions alone may not be sufficient to 83 distinguish competing hypotheses [15, 16]. A more explicit description of the processes 84 underlying all hypotheses will generate a wider range of predictions which can be used to 85 86 disentangle possibly non-mutually exclusive hypotheses and evaluate the relative importance of these processes. 87

88

89 We, therefore, call for a transformation in the way biologists think about and study the LDG. The classification of hypotheses (Table 1) is an important first step, but it does not resolve the 90 91 difficulty of identifying and quantifying the relative strength of the processes underlying the LDG. We propose moving towards a mechanistic framework, founded on key processes that 92 93 describe how individual organisms interact with their biotic and abiotic environments, and 94 how these interactions scale up to result in the LDG and other secondary biodiversity patterns. Ultimately, revealing the nature of these eco-evolutionary processes will yield 95 more insight than continuing to argue about non-mutually exclusive LDG hypotheses. 96

97

98 Examining the LDG through the lens of mechanistic macroecology

99 Key processes across levels of biological organization

100 We recognize four key processes, as defined by [17], that necessarily underpin the LDG and

- thus should be included as components of any LDG model that aims to capture variation in
- species richness, abundance, and composition over a spatially and temporally variable

environment: 1) selection, 2) ecological drift, 3) dispersal, and 4) speciation. Selection, drift, 103 104 and dispersal can all influence the birth, death, and movement of individuals over small spatial and temporal scales. Selection (sensus [17]) encompasses any process that results in 105 the differential survival and reproduction of individuals, based on how environmental 106 filtering [18] and biotic interactions select for specific traits. Ecological drift manifests itself 107 via stochastic variation in the births and deaths of individuals. Dispersal of individuals is 108 109 influenced by the spatial structure of the landscape as well as individual dispersal capabilities and can lead to species colonizing new regions. Each of these individual-level ecological and 110 microevolutionary processes is propagated throughout higher levels of biological 111 112 organization, resulting in discrete patterns at the level of populations, species, and 113 communities (Figure 1).

114

Over longer timescales, environmental conditions have fluctuated with glacial/interglacial 115 oscillations, cooler and warmer periods in Earth's history, orogenic events, volcanic activity, 116 117 and shifts in tectonic plates, all of which can affect diversity dynamics [19-21]. At these spatial and temporal scales selection, ecological drift, and dispersal determine where species 118 119 or even whole clades are able to persist geographically and how traits evolve. Species that 120 become poorly adapted to the environment or that are poor competitors for resources are expected to have low fitness and ultimately go extinct, reflecting critical eco-evolutionary 121 feedbacks [22, 23]. Speciation becomes especially relevant with increasing temporal and 122 123 spatial scales. The details of how speciation occurs are complex and the critical question in a LDG context becomes how and why speciation mode or rate varies along geographic 124 gradients. All of the processes described above necessarily interact with each other and with 125 the spatiotemporal environment, resulting in the broad range of geographic and phylogenetic 126 biodiversity patterns that we observe today. As highlighted below, these processes can help us 127 compare and disentangle LDG hypotheses. 128

130 Classical LDG hypotheses revisited

Characterizing LDG hypotheses based on the key processes described above helps to clarify 131 the internal logic of those hypotheses, and highlights how they differ. All hypotheses invoke 132 an explicit driver or condition that varies latitudinally (Figure 1), but considering the 133 processes related to this driver, often below the level of biological organization at which the 134 hypothesis was formulated, can reveal previously unrecognized assumptions and predictions. 135 Below we discuss four examples, chosen to represent hypotheses invoking variation in limits, 136 rates, and time. These examples may also serve as a guide for better understanding other 137 138 hypotheses.

139

140 *The more individuals hypothesis*

141 The "more individuals hypothesis" invokes latitudinal variation in ecological limits and a positive relationship between the number of species and resource availability [24]. If 142 143 resources are finite and a zero-sum constraint on the total amount of biomass or individuals applies, any increase in diversity over time results in a decrease in average biomass or 144 abundance per species. Extinction rates will thus be diversity-dependent and richness will be 145 146 regulated around some equilibrial value that scales with the total number of individuals that can be supported [24, 25]. This hypothesis implicitly invokes interspecific competition and 147 the resultant allocation of resources across species (Table 1). The argument does not invoke 148 selection (Fig. 1) and can be applied equally to ecologically neutral or non-neutral species. An 149 important and unstated assumption is that the response of the biota to environmental change is 150 fast enough that richness is at equilibrium across the latitudinal gradient. 151

152

153 The seasonality hypothesis

The seasonality hypothesis argues that the within-year environmental stability of the tropics 154 155 results in either greater diversification rates or higher ecological limits via increased niche packing (Table 1 and Fig. 1). The first argument suggests that in the less seasonal tropics, 156 organisms experience a smaller range of conditions and hence evolve narrower thermal niches 157 compared to the temperate zone. The idea that "mountain passes are higher in the tropics" 158 [26] suggests that dispersal barriers were effectively greater there, increasing the chance of 159 160 population divergence and allopatric speciation [27, 28]. Selection thus dictates the environmental conditions that a species can tolerate, but it is speciation rate that varies with 161 latitude and ultimately generates the LDG. The second version of the seasonality hypothesis 162 163 suggests that stability-driven specialization promotes intense niche packing, and hence more species can coexist in the tropics [29, 30]. Species are then hypothesized to evolve narrower 164 resource breadths rather than narrow thermal niches, assuming that resources are limited and 165 166 that diversity actually emerges from niche packing [29] (Table 1, Figure 1). Implicit in both hypotheses is a performance tradeoff between specialists and generalists, such that specialists 167 evolve and outcompete generalists in aseasonal environments. 168

169

170 The temperature-dependent speciation rates hypothesis

171 The hypothesis that higher temperature elevates evolutionary rates has been used to explain global diversity patterns in both land and sea [31, 32]. One version of the hypothesis [33] 172 follows from the metabolic theory of ecology [34], stating that temperature positively affects 173 all biological rates including mutation rates, which can lead to speciation and ultimately 174 diversity accumulation. This assumes that speciation rates directly follow from mutation rates, 175 which may be problematic if other factors (e.g. existence of geographic barriers, assortative 176 mating) are limiting speciation. The hypothesis makes no specific predictions regarding 177 selection or dispersal. Importantly, this hypothesis could be invoked in either an equilibrium 178 or non-equilibrium world. In a non-equilibrium world, speciation rates alone could explain 179

variation in richness between regions if all regions were similarly old, and extinction rates
were equal across regions [10]. In an equilibrium world, increased speciation rates in the
tropics can lead to higher equilibrium richness, as in Hubbell [35] neutral model of
biodiversity.

184

185 The tropical niche conservatism hypothesis

The tropical niche conservatism hypothesis [36, 37] states that diversity is higher in the 186 tropics because of the infrequency of colonisations of the cooler temperate zone by a tropical 187 ancestor due to strongly conserved thermal niches and tropical origins of most taxa, and hence 188 189 the longer time available for diversification in the tropics. The hypothesis assumes that, barring major disturbances or climatic shifts, species richness will continue to increase 190 unbounded over time [37]. This hypothesis has only ever been formulated at the species level, 191 192 and yet it inherently implies a particular set of rules by which individuals interact with the environment and each other. Selection by the environment is by definition strong, with 193 194 individuals unable to survive and reproduce under conditions different from their optima, and evolution of a new optimum is rare. Less obvious are the implications of the hypothesis for 195 196 resource competition between individuals. Unbounded, or diversity-independent, 197 diversification is only possible in the absence of an overarching zero-sum constraint [25]. The absence of such a constraint implies that while the population size of a species might be 198 affected by the fit between the environment and environmental performance traits, it is 199 200 independent of the population sizes of potential competitors and of interspecific competition more broadly. 201

202

203 The utility of a mechanistic framework

The examples presented above illustrate three insights gained by adopting a generalized eco-evolutionary framework. First, many of the fundamental rules by which organisms are

assumed to interact with each other and with their environment will be qualitatively similar 206 207 regardless of LDG hypothesis. For example, individual survival and reproduction must be functions of how well adapted the individuals are to their environment relative to their intra-208 and interspecific competitors. Second, latitudinal differences in ecological limits, 209 diversification rates, and time for diversification may emerge via different mechanisms 210 integrated into the same framework. For example, diversification rates may be higher due to 211 212 the temperature-dependence of mutation rates [9, 38] or due to the increased reproductive isolation in aseasonal environments [27, 39]. Third, although each hypothesis invokes a 213 primary driver or process, we have shown that these hypotheses also make unstated 214 215 assumptions about other processes and mechanisms which need to be considered in concert to fully understand the emergence of the LDG and other macroecological and macroevolutionary 216 217 patterns.

218

219 Mechanistic eco-evolutionary models as a quantitative tool for

220 understanding LDG patterns

The mechanistic framing of processes that underpin the LDG naturally facilitates the 221 222 translation from heuristic thinking to mechanistic eco-evolutionary models (Box 1). We believe that building these models will be essential to making progress on the LDG and 223 224 biodiversity patterns in general because they allow quantitative analyses and predictions of the various secondary patterns. Secondary patterns are key for more powerful inference about 225 the origin of species richness patterns. Below we provide concrete examples of components of 226 a mechanistic LDG model and associated patterns followed by a discussion about how to use 227 228 such a model for inference with the available data.

229

230 Mechanistic models for studying the LDG

231 *The spatiotemporal environmental template*

The basic driver of an LDG model is the spatiotemporal environmental template. It can be viewed as the theater in which the eco-evolutionary play unfolds, and the spatiotemporal variation in that template (Earth's climatic, geologic, and tectonic history) may be as critical to emergent diversity patterns as the mechanisms and processes governing how organisms interact and evolve [40-42]. Explaining the LDG with eco-evolutionary **simulation models**, therefore, benefits from suitable paleoenvironmental reconstructions [43] and the integration of global data sets on continental topography and paleoshorelines [44, 45].

239

240 Trait-based local population dynamics

Traits are essential for individual survival and reproduction (fitness) and mechanistic models 241 that include interactions of organismal traits and the abiotic and biotic environment, below the 242 243 level of species (i.e. at the individual, population or metapopulational level), is thus appropriate. Local population dynamics can, for example be assumed to be trait-dependent 244 245 [46, 47]. One set of traits might determine an organism's fitness dictated by the abiotic environment, a different set of traits may influence relative fitness associated with the suite of 246 247 potential competitors present at any point in time [48]. Such a modeling approach requires 248 making basic assumptions that facilitate the link between environmental conditions, available resources, and ecological interactions, and population dynamics then emerge from those 249 assumptions. 250

251

252 Spatial and eco-evolutionary metacommunity dynamics

For modeling eco-evolutionary metacommunity dynamics, trait-based models need to be implemented in a larger spatial context, allowing individuals to disperse over geographically relevant extents. Metacommunity dynamics will arise from eco-evolutionary feedbacks between dispersing individuals and recipient communities within the context of the

spatiotemporal template [49]. Evolutionary dynamics result from natural selection by both 257 258 abiotic and biotic conditions, ecological drift, dispersal, and speciation. Speciation can be modeled using a phenomenological approach or more complex allele-based models in which 259 phenotypic trait variability is completely or partially heritable and the accumulation of genetic 260 incompatibilities may drive differentiation of daughter species (Box 2). Each of these 261 modeling components is necessary for capturing the suite of processes invoked by LDG 262 hypotheses (Box 1), they can be modeled with varying degrees of complexity and they come 263 with a set of low-level assumptions that need to be clearly stated (Box 2). 264

265

266 Understanding patterns and inferring processes

Above we have shown that a mechanistic mindset is useful to better understand the internal 267 logic and consequences of the different hypotheses, as well as the interactions among them. In 268 269 addition, a mechanistic model can clarify the biodiversity patterns expected under different combinations of spatiotemporal environmental templates, biotic interactions, and other eco-270 271 evolutionary rules [e.g. 16, 48, 50]. This ability to simulate very different worldviews of how the LDG arises (e.g., "ecological limits", "niche conservatism", etc.) within the same 272 comparative framework is a critical element of our approach as different types of processes 273 274 modeled with varying degrees of mechanistic detail can be explored and contrasted.

275

Ultimately, we need mechanistic models to understand the details of the emerging ecoevolutionary patterns at a sufficient resolution to be able to quantitatively confront them with
data. The more secondary patterns (e.g. phylogenies, species ranges, distributions of
abundance or functional traits) that can be modeled, the greater the diagnostic power of the
model for exploring parameter space and for inferring the strength and interactions of
different processes. The examination of these patterns will also point to the type of data that
will be most valuable for reliable inference of a given process [51].

While we believe that confronting different model scenarios with multiple observed patterns 284 (described in Box 3) is the only way to make progress in understanding the LDG, we realize 285 that substantial conceptual, statistical, and computational challenges are associated with this 286 task [52]. The complexity of the suggested models often makes it difficult to understand the 287 consequences of the underlying assumptions. Ways of overcoming such challenges are to 288 build on known ecological models (e.g. Lotka-Volterra equations) and evolutionary theory 289 (e.g. Adaptive Dynamics theory) that has been studied extensively. The models should also be 290 built and analyzed in a sequential manner of increased complexity to shed light on the 291 292 consequences of the key model assumptions and their interactions. While it is not our aim to detail these and other methodological challenges here, we nevertheless highlight two basic 293 inferential approaches that seem particularly promising. First, qualitative matching of multiple 294 295 patterns gives an indication of whether the modeled processes can produce the patterns that we observe [15, 25, 41, 53]. Pattern matching is conceptually straightforward and easily 296 297 allows combining the LDG with multiple observed secondary patterns to compare alternative model or parameter choices. Second, models like the ones suggested above can be fitted to a 298 299 range of patterns in data using simulation-based methods such as Approximate Bayesian 300 **Computation** [54-57] or synthetic likelihood [58, 59]. Regardless of which inferential approach is used, any empirical patterns that a model is unable to reproduce can be instructive 301 in the iterative process of model improvement. 302

303

304 Concluding remarks

Progress in understanding the processes that underlie LDG patterns and associated diversity
patterns has been slow (see also Outstanding Questions). We repeat calls for a transition in
biodiversity research, translating verbal models into a unified mechanistic framework that can
be implemented in quantitative computer simulations [52, 53, 60]. In such a framework,

researchers can focus on measuring and inferring the ecological and evolutionary processes 309 310 that govern the interaction of organisms with each other and their environment in time and space, which must ultimately underpin the LDG. By applying this framework, hidden 311 assumptions in current hypotheses are exposed, revealing how the hypotheses relate to each 312 other and how they might be distinguished (Table 1, Figure 1). More importantly, this 313 framework is a roadmap for flexible eco-evolutionary simulation models (Box 1-2) that can 314 315 generate a rich set of empirical patterns from the same underlying processes. We believe that this ability to produce multiple diagnostic patterns will be crucial for inference (Box 3), and 316 ultimately for converting the available data into new knowledge about macroecology and 317 318 macroevolution. Challenges associated with model construction and the way models are confronted with data will arise, but such challenges are inherent and inevitable to all sciences 319 that deal with complex systems. We are confident that, with time, these challenges can be 320 321 addressed, and models combining realistic spatiotemporal environmental templates with trait based eco-evolutionary implementation under an iterative procedure of model design, 322 323 evaluation and improvement, will advance our understanding and quantitative inference of the processes underlying the LDG. 324

325

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