

1 **Title:** Environmental variables that influence resource temporal variability and abundance drive  
2 trophic diversity in mammals

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34 **Abstract**

35 Understanding environmental drivers of species diversity has become increasingly  
36 important under climate change. Different trophic groups (predators, omnivores, herbivores)  
37 interact with their environments in fundamentally different ways, and may therefore be  
38 influenced by different environmental drivers. Using random forest models, we identified drivers  
39 of terrestrial mammals' total and proportional species richness within trophic groups at a global  
40 scale. Precipitation seasonality was the most important predictor of richness for all trophic  
41 groups. Richness peaked at intermediate precipitation seasonality, indicating that moderate levels  
42 of environmental heterogeneity promote mammal richness. Gross primary production (GPP) was  
43 the most important correlate of the relative contribution of each trophic group to total species  
44 richness. The strong relationship with GPP demonstrates that basal-level resource availability  
45 influences how diversity is structured among trophic groups. Our findings suggest that  
46 environmental characteristics that influence resource temporal variability and abundance are  
47 important predictors of terrestrial mammal richness at a global scale.

48 **Introduction**

49 Global geographic patterns in species richness have been studied since the 19<sup>th</sup> century  
50 (Hawkins 2001). Although we have posited a host of explanations, there is no consensus  
51 regarding the mechanisms that drive geographic patterns of biodiversity. This lack of consensus  
52 is perhaps because no single mechanism dictates diversity at such a broad geographic scale, and  
53 because different drivers may underly patterns for different organismal groups (Richardson &  
54 Pennington 2016).

55 Several mechanisms hypothesized to underpin the global richness gradient are related to  
56 energy and resource acquisition, which are also integral to understanding ecological food webs.  
57 Despite the overlap between diversity and food web theories, how diversity gradients vary

58 among different trophic groups (i.e., herbivores, omnivores, and predators) has only been  
59 described at the global scale for arthropods (Castagneyrol & Jactel 2012) and coral reef fishes  
60 (Siqueira *et al.* 2023), and at a very coarse spatial scale for the role of trophic interactions in  
61 shaping mammalian diversity patterns (Zhang *et al.* 2018a). Our study helps identify drivers that  
62 underly species richness patterns in different mammalian trophic groups, shedding new light on  
63 drivers of ecological patterns in species and providing valuable information for biodiversity  
64 conservation in an era of global change.

65 Broadly, two major categories of mechanisms have been invoked to explain latitudinal  
66 patterns of species richness: evolutionary mechanisms and ecological mechanisms. Evolutionary  
67 mechanisms focus on the processes that promote speciation and extinction, whereas ecological  
68 mechanisms focus on the current and historical abiotic and biotic interactions that help shape  
69 diversity patterns (Mittelbach *et al.* 2007). Although evolutionary processes are ultimately  
70 responsible for creating diversity, ecological conditions can affect speciation (e.g., diversification  
71 rate), extinction, and dispersal processes to influence spatial patterns in species and trait-based  
72 richness (Graham *et al.* 2014; Wiens 2023). Furthermore, because diversity is rapidly declining  
73 in response to changing ecological conditions, identifying the ecological mechanisms that help  
74 drive richness patterns may provide better insight for managing and predicting diversity under  
75 global change.

76 Recent increases in the availability of animal diet and climate data now allow scientists to  
77 examine patterns and ecological drivers in an understudied area of diversity, trophic diversity. In  
78 this study, we focus on three longstanding ecological-diversity hypotheses that can be tested  
79 using global-scale climate data to explain patterns in trophic richness in mammals: the species-

80 energy hypothesis (Willig *et al.* 2003), the environmental heterogeneity hypothesis (Stein *et al.*  
81 2014), and the environmental harshness hypothesis (Chesson & Huntly 1997).

82 The species-energy hypothesis posits that species richness increases with the amount of  
83 energy and resources available to species (Willig *et al.* 2003) and has previously been assessed  
84 using plant productivity [e.g., gross primary production (GPP), net primary production (NPP),  
85 normalized difference vegetation index (NDVI)], solar radiation, or temperature data. The  
86 species-energy hypothesis implies that richness increases with the quantity and quality of  
87 biomass available lower on the food web, and species richness should decrease sequentially with  
88 increasing trophic levels (Hawkins *et al.* 2003). Therefore, if the species-energy hypothesis  
89 holds, herbivore richness would be more tightly coupled to plant productivity than other trophic  
90 groups due to their position on the food web. Studies have demonstrated that, among arthropods,  
91 positive relationships between plant and consumer diversity are dampened at higher trophic  
92 positions (Haddad *et al.* 2009; Scherber *et al.* 2010). This effect could be due to a positive  
93 relationship between plant diversity and productivity, leading to greater resource availability for  
94 herbivores (Tilman *et al.* 1996, Zhang *et al.* 2018b). A second, non-exclusive explanation is that  
95 plant diversity increases the number of ecological niches for all trophic groups by increasing  
96 habitat complexity (Castagneyrol & Jactel 2012).

97 The environmental heterogeneity hypothesis suggests that environmental variability (here  
98 proxied by precipitation seasonality, temperature seasonality, and isothermality) affects diversity  
99 by influencing niche specialization and competition. Environmental heterogeneity operates on  
100 both a spatial and temporal scale, and climate data lends itself particularly well to testing the  
101 temporal component. The impact of temporal environmental fluctuations on diversity can be

102 tested on a daily, seasonal, or annual scale using temperature and precipitation data, two of the  
103 most important environmental filters that influence terrestrial species distributions.

104 Predictions about the direction of the effects of heterogeneity on richness are varied.

105 Some have suggested that stable environments promote greater diversity through increased  
106 specialization, as species can focus on a narrow, but predictable resource to reduce competition.

107 Meanwhile, unstable environments could limit diversity by imposing physiological constraints

108 on organisms (Klopf 1959; Luo *et al.* 2012). Conversely, others have posited that some

109 environmental fluctuation or disturbance promotes species co-existence by preventing

110 competitive exclusion (e.g., the intermediate disturbance hypothesis; Hutchinson 1961; Connell

111 1978). This dichotomy of the impact of environmental stability on species richness may be due

112 to the temporal scale considered. For example, short-term (i.e., seasonal) environmental

113 fluctuations may promote biodiversity by allowing for predictable variation (i.e., variability that

114 recurs reliably), whereas longer-term fluctuations may negatively impact biodiversity by forcing

115 species to endure sub-optimal conditions for extended periods (Liu *et al.* 2021).

116 Due to their generalist lifestyle, omnivores may better tolerate fluctuations in

117 environmental conditions and food availability due to flexibility in the types of resources they

118 exploit, and may rely on environmental fluctuations to balance competitive abilities (Wootton

119 2017). Environmental fluctuations that alter food quality and quantity may improve omnivores'

120 fitness relative to predators and herbivores because they can change their feeding habits to

121 tolerate a wider range of environmental conditions (Kondoh 2008). For example, omnivorous

122 fish in environments with fluctuating water levels consume different amounts of plant vs. animal

123 tissue during wet seasons compared to dry seasons (McMeans *et al.* 2019). In contrast, in stable

124 systems, omnivores are often inferior competitors compared to herbivores and predators due to  
125 trade-offs associated with maintaining the ability to consume multiple food types (Diehl 2003).

126 The environmental harshness hypothesis also predicts that the environment operates on  
127 diversity by influencing niche differentiation and competition (Chesson & Huntly 1997), and  
128 could be assessed using extremes in mean annual temperature and precipitation. Similar to the  
129 effects of environmental variability on trophic structure diversity, omnivores may be more  
130 tolerant of harsh environments than herbivores and predators. For example, omnivores can adjust  
131 their degree of herbivory or carnivory in response to temperature-induced changes in food  
132 quality and quantity (Boersma *et al.* 2016; Zhang *et al.* 2018b).

133 This study aims to determine whether the three ecological mechanisms outlined above  
134 operate differently at different trophic levels of mammals. Here, we posit that global patterns of  
135 mammal species richness will differ among trophic groups (herbivores, omnivores, and  
136 predators) due to different dominant environmental mechanisms influencing their diversity.

137 Specifically, we hypothesize that 1) plant production will be the dominant driver of herbivore  
138 diversity and be less important for other trophic groups, and 2) that the negative impacts of  
139 environmental harshness and temporal heterogeneity will be greater for herbivores and predators  
140 than omnivores. We assessed these hypotheses by leveraging multiple global-scale, spatial  
141 datasets that characterize mammal species richness and ecological properties.

## 142 **Methods**

### 143 *Trophic Categories and Environmental Variables*

144 We used and updated a database by Atwood *et al.* (2020) to assess global, extant  
145 terrestrial mammal biodiversity across trophic levels. The original database contained trophic  
146 group classifications (predator, omnivore, and herbivore) and geographical data on all extant

147 mammals assessed on the 2019-2 IUCN Red List of Threatened Species (IUCN 2019). We  
148 augmented the database by adding previously unincluded extant mammals assessed on the 2021-  
149 1 IUCN Red List of Threatened Species (IUCN 2021). We characterized the diets and trophic  
150 groups of species following an approach described in detail by Atwood *et al.* (2020). Briefly, we  
151 searched published literature and reference texts for diet information from wild populations (see  
152 supplementary table 1 for diet source references). For ~5% of species, diet information was not  
153 available in published resources; for these species, we extrapolated a diet from the most closely  
154 related taxonomic group with published diet information, typically congeners or confamilials.  
155 Atwood *et al.* (2020) used this extrapolation approach in their study and found it predicted coarse  
156 trophic groups of mammals with 94% accuracy. We classified species as either predators,  
157 omnivores, or herbivores based on the contribution of plant vs. animal material to a species' diet:  
158 we defined predators as species that consume  $\geq 80\%$  animal-based diets, herbivores as species  
159 that consume  $\geq 80\%$  plant-based diets, and all other species as omnivores.

160 To reduce overall uncertainty in diet classifications and trophic diversity patterns, we did  
161 not classify species' diets further than predator, herbivore, or omnivore. Past quantifications of  
162 more refined diet categories, such as those in EltonTraits 1.0, only provide semiquantitative  
163 information on the importance of different diet categories, and taxonomic extrapolations were  
164 used to fill in missing species' diets (Wilman *et al.* 2014; Atwood *et al.* 2020). While  
165 extrapolating diets from related taxa performs well for coarse diet categories (i.e., herbivore,  
166 omnivore, predator), it is less accurate for more refined diets (Gainsbury *et al.* 2018). Finally,  
167 grouping species into more refined diet categories (frugivore, insectivore, etc.) can obscure  
168 global diversity patterns because, unlike coarse categorizations, most species are not confined to  
169 a single category, and diets may vary spatially, seasonally, and temporally for a single species.

170 Without quantitative information that captures spatial and temporal variation in the importance  
171 of different diet items to a species, it becomes difficult to ascertain global patterns and their  
172 drivers for more refined diet categories.

173 We used gross GPP as an indicator of plant production and resource availability. We used  
174 isothermality and seasonal variations in temperature and precipitation as proxies for temporal  
175 environmental heterogeneity. Finally, we characterized environmental harshness based on mean  
176 annual temperature (MAT) and mean annual precipitation (MAP).

177 We leveraged published geographic environmental databases as sources for  
178 environmental variables. We extracted MAT ( $^{\circ}\text{C}$ ), temperature seasonality (standard deviation  $\times$   
179 100), isothermality (diurnal range/annual range), MAP (cm), and precipitation seasonality  
180 (coefficient of variation) from the WorldClim v. 2 global climate database at a resolution of 30  
181 arc seconds (0.083 degrees;  $\sim 85 \text{ km}^2$ ); values were annual means from 1970-2000 (Fick &  
182 Hijmans 2017). We used annual GPP ( $\text{g C m}^{-2} \text{ y}^{-1}$ ) at a 0.05 degree resolution ( $\sim 30 \text{ km}^2$ ) (Zhang  
183 *et al.* 2017).

184 There has been great debate about the implications of spatial scale on species distribution  
185 data, with some arguing that analyses using a finer resolution than 200 km  $\times$  200 km  
186 overestimates occupancies (Hurlbert & Jetz 2007). Yet others have suggested that such a coarse  
187 grain resolution degrades the data and obscures important diversity patterns driven by finer-scale  
188 changes in climatic variables, such as those imposed by changes in elevation (Jenkins *et al.*  
189 2013). Acknowledging that some species distributions could be overestimated, we gridded the  
190 terrestrial surface of the globe into 30 km  $\times$  30 km pixels to better capture relevant variability in  
191 climatic variables. We calculated the average of each environmental variable, total mammal  
192 species richness, species richness for each trophic group within each pixel. We also calculated

193 the relative contribution of each trophic group to total richness in each pixel, and we identified  
194 pixels where a particular trophic group is overrepresented. We considered a trophic group to be  
195 overrepresented if its relative contribution to total diversity in each pixel was greater than its  
196 global upper quartile value. For example, the global upper quartile for relative predator richness  
197 is 52.17%. Thus, they are overrepresented in pixels where they contribute to more than 52.17%  
198 of relative diversity. In this interpretation, overrepresentation indicates that a geographic region  
199 is more favorable to a certain trophic lifestyle than others.

200 *Random Forest Modelling*

201 We used random forest modeling to identify environmental drivers of total mammal  
202 richness, the richness of each trophic group, and the relative contribution of each trophic group  
203 to total richness. We opted to use random forest modeling due to its robustness in dealing with  
204 non-linear responses and because it allowed us to determine the relative importance of different  
205 environmental variables (Cutler *et al.* 2007). To balance the distribution of mammal richness for  
206 our analyses, we performed stratified sampling of  $30 \times 30$  km geographic pixels based on total  
207 mammal richness. We grouped pixels into 19 bins at richness intervals of ten (i.e., 0-10 species,  
208 10-20 species, etc.) except for richness levels greater than 180 (max=228), which we grouped  
209 into a single bin. We randomly sampled 1500 pixels from each bin for a total of 28,500 pixels.  
210 Our stratified sampling approach also mitigates spatial autocorrelation, reducing the number of  
211 adjacent or nearby pixels used in the data set (Chevalier *et al.* 2021).

212 Each random forest model initially included latitude coordinate, MAT, temperature  
213 seasonality, isothermality, MAP, precipitation seasonality, and GPP as explanatory variables. We  
214 performed reverse-fold cross-validation to determine the optimal number of variables for each  
215 model and eliminated non-useful variables based on a measure of variable importance (increase

216 in MSE) and on variance inflation factors (VIF). We assessed the marginal effects of each  
217 predictor variable using random forest partial dependency plots (Breiman 2001). Random forest  
218 models were constructed using the *randomForest* package (Liaw & Wiener 2002) in the R  
219 statistical computing environment (R Core Team 2022).

220 **Results**

221 Our final dataset included 5713 unique mammal species, with 2024 species classified as  
222 predators, 2268 species classified as herbivores, and 1421 classified as omnivores. The global  
223 average of terrestrial mammal richness is  $59.41 \pm 44.24$  (SD) species per  $30 \times 30$  km pixel. Mean  
224 per pixel predator richness is  $28.57 \pm 22.31$  species, mean herbivore richness is  $18.13 \pm 14.40$   
225 species, and mean omnivore richness is  $12.72 \pm 9.34$  species. Predator, herbivore, and omnivore  
226 mammal diversity all peak near the equator and decrease poleward (Figs. 1 and 2a). Across most  
227 regions of the globe, predator richness is higher than the other trophic groups, followed by  
228 herbivore richness and omnivore richness. However, from  $\sim 15\text{--}30$  °N, omnivore richness is  
229 slightly higher than herbivore richness. At the northernmost latitudes ( $>65$  °N), herbivore  
230 richness tends to be similar or slightly higher than predator richness (Fig. 2). On average,  
231 predator species represent  $46.75 \pm 11.93\%$  of per pixel richness. In contrast, herbivore species  
232 represent  $30.16 \pm 11.64\%$ , and omnivore species represent  $22.72 \pm 9.69\%$ . The relative  
233 contribution of each trophic group to total richness varies with latitude (Fig. 2b). Predator  
234 relative richness exhibits local maxima at  $\sim 25$  °S and  $\sim 10$  °N. Herbivore relative richness  
235 exhibits local maxima at  $\sim 5$  °S and  $\sim 60$  °N, and omnivore relative richness exhibits local  
236 maxima at  $\sim 10$  °S and  $\sim 25$  °N.

237 Each trophic group exhibits different patterns of over-representation across the globe  
238 (Fig. 3). Over much of the tropics and subtropics, omnivory is overrepresented. In the temperate

239 zones, particularly the northernmost parts of the temperate zone, herbivory tends to be  
240 overrepresented. Predators do not exhibit clear latitudinal patterns in overrepresentation, but are  
241 overrepresented in portions of tropical Africa, temperate humid regions of Europe and northern  
242 Africa, and Greenland tundra (Fig. 3).

243 *Random Forest Models – Raw Species Richness*

244 Based on reverse fold cross-validation and percent increase in mean square error  
245 (increase in MSE), we included the following five parameters in all trophic models: precipitation  
246 seasonality, latitude coordinate, GPP, MAP, and MAT (Table 1). We omitted isothermality and  
247 temperature seasonality because they were the least important variables based on increase in  
248 MSE. They also exhibited the highest VIFs at 6.82 and 8.90, respectively. All random forest  
249 models explained high amounts of variance, ranging from 87.29-91.60% variance explained.

250 Random forest modeling revealed commonalities in predictors of total mammal richness  
251 and richness within trophic groups. For total mammal richness and all three trophic groups, an  
252 increase in MSE indicates that precipitation seasonality is, by a large margin, the most important  
253 variable explaining global richness patterns (Table 1). Latitude is the second most important  
254 explanatory variable for total mammal and herbivore richness. In contrast, GPP is a more  
255 important explanatory variable for predator richness than latitude. For omnivore richness, MAT  
256 and MAP are more important than latitude.

257 Although the relative importance of the explanatory variables differed among trophic  
258 groups, partial dependency plots indicate that species richness of different trophic groups tends  
259 to respond similarly to the explanatory variables, except for the response to MAT (Fig 4).  
260 Predator richness increases with MAT to ~15 °C and then exhibits fairly consistent diversity  
261 levels at higher temperatures. Herbivore richness also increases to a maximum at ~15 °C,

262 whereas omnivore richness achieves a maximum at a lower temperature of  $\sim 12$  °C. In contrast to  
263 predator richness, herbivore and omnivore richness decrease at higher temperatures.

264 *Random Forest Models – Relative Contribution to Richness*

265 Because patterns in species richness across trophic levels can be masked by the  
266 disproportionate number of species in different regions, we also investigated the mechanisms  
267 controlling diversity across different trophic groups by using their relative contribution to  
268 richness. Reverse-fold cross-validation indicated that the best model configuration included five  
269 explanatory variables. However, variable importance measures based on increase in MSE  
270 indicated different top predictors for the three trophic groups. To facilitate comparisons among  
271 trophic groups, we retained the six most important variables in all models (GPP, precipitation  
272 seasonality, temperature seasonality, MAT, MAP, and latitude coordinate) according to their  
273 increase in MSE (Table 2). We omitted isothermality because it was the least important variable  
274 in all models. For predators and herbivores, GPP was the most important predictor explaining the  
275 relative contributions of trophic groups to total richness. Precipitation seasonality and  
276 temperature seasonality were the next most important predictors for predator relative richness  
277 and were of similar importance. Temperature seasonality, MAP, and latitude were the next most  
278 important variables explaining herbivore relative richness and were all of similar importance.  
279 The top three most important variables for omnivore relative richness (precipitation seasonality,  
280 MAP, and GPP) were equally important. The individual models for predator, omnivore, and  
281 herbivore relative richness explained 79.91%, 73.40%, and 76.30% of the variance, respectively.

282 Partial dependency plots indicate that for any given variable, trophic groups varied in  
283 their response (Fig. 4). Here, we focus on the responses of relative richness to GPP and  
284 precipitation seasonality, which tended to be among the most important predictors for all trophic

285 groups (Table 2). The relative richness of omnivores was negatively related to GPP between  
286 100-1500 g C m<sup>-2</sup> y<sup>-1</sup>, decreasing by ~5% between these values. In contrast, relative richness for  
287 predators and herbivores increased by ~3.5% and ~1%, respectively, between these GPP values.  
288 At GPP values above 1500 g C m<sup>-2</sup> y<sup>-1</sup>, the relative richness of all three trophic groups remained  
289 fairly consistent. Predator relative richness was positively related to precipitation seasonality  
290 between 0-125 (CV), increasing by ~4.5% between these values. Omnivore relative richness  
291 decreased by ~4% between the same precipitation values, while herbivore relative richness  
292 decreased by ~0.5%.

293 **Discussion**

294 The objective of our study was to determine whether environmental conditions  
295 differentially impact mammal species richness at different trophic levels. Our results support the  
296 hypothesis that species richness spatially varies across trophic groups, with mammalian  
297 predators, herbivores, and omnivores dominating different regions of the globe. A combination  
298 of climate and productivity variables helps predict the global patterns in total mammal richness,  
299 richness within trophic groups, and the relative contribution of trophic groups to total richness.  
300 However, the relative importance of different environmental characteristics and their relationship  
301 with richness varied among predators, omnivores, and herbivores. These differences suggest that  
302 the interplay between a mammal's environment and its trophic strategy influences how  
303 mammalian diversity is structured.

304 We did not find support for our hypothesis that GPP would be a more important predictor  
305 for herbivores and omnivores than predators. Although GPP emerged as an important predictor  
306 of total mammal richness, when segregated by trophic groups, it was a more important predictor  
307 for total predator richness than herbivore or omnivore richness. Varied responses by the different

308 trophic groups to GPP indicate that future changes in this variable could alter how diversity is  
309 structured among trophic levels, with predators experiencing the most dramatic effects with  
310 changes to GPP.

311 When considering the relative contribution of trophic groups to diversity, GPP was again  
312 an important predictor for predators, herbivores, and omnivores. However, in contrast to  
313 predators and herbivores, which increased in proportion with GPP, the proportion of omnivores  
314 slightly declined with increasing GPP. Studies on the tempo of lineage diversification and  
315 trophic transition in mammals show that omnivore diversity primarily evolves through  
316 transitions into that strategy from herbivores and carnivores (Price *et al.* 2012). Additionally,  
317 herbivores and carnivores have developed greater diversity than omnivores through  
318 specialization and subdivision of niches (Price *et al.* 2012). Thus, one might expect that under  
319 higher GPP, more specialized trophic niches would develop for herbivores and, subsequently,  
320 carnivores, suppressing the need to transition to omnivory. Simulation models support this  
321 hypothesis by showing that the evolution of omnivory decreases with increased plant production  
322 (Chubaty *et al.* 2014).

323 Our findings that GPP is the most important variable predicting the relative contribution  
324 of trophic groups to total diversity and the generally positive relationship between GPP and  
325 species richness for all trophic groups indicates that total energy and basal resource availability  
326 are key correlates of diversity. These results support the species-energy hypothesis, which posits  
327 that overall diversity is positively correlated with the energy available in an ecosystem (Wright  
328 1983). Our finding that the magnitude of the effect of GPP on species diversity is greatest at the  
329 highest trophic level (i.e., for predators) provides an interesting new context to the version of the  
330 species-energy hypothesis asserting that productivity exerts bottom-up controls on diversity.

331        The bottom-up formulation of the species-energy hypothesis suggests that the number of  
332    trophic levels is regulated by the energy available at the base of the food chain, and that in areas  
333    of high productivity, longer food chains promote greater predator richness (Evans *et al.* 2005a,  
334    b). However, in this explanation, herbivores and omnivores are expected to have a stronger  
335    positive relationship with productivity than predators (Hawkins *et al.* 2003; Jetz *et al.* 2009). Yet,  
336    we observed the opposite: the rate of increase of species richness with GPP is greater for  
337    predators than for herbivores and omnivores. Furthermore, when considering simple bivariate  
338    correlations between GPP and predator, herbivore, or omnivore richness in our global dataset  
339    ( $n=134,491$ ), GPP is more strongly correlated with predator richness (Spearman's  $\rho=0.74$ ) than  
340    herbivore or omnivore richness ( $\rho=0.70$  and  $\rho=0.65$ , respectively). Together, this indicates that  
341    the influence of GPP on mammal diversity is not necessarily damped in higher trophic groups.  
342    However, our findings do not entirely refute the bottom-up hypothesis. Although we did not  
343    assess diet specialization within trophic groups in this study (i.e., classifying mammals as  
344    insectivores, frugivores, granivores, etc.), Atwood *et al.* (2020) found that the diets of >90% of  
345    extant mammalian predators included insects. Thus, we cannot conclude that predator diversity is  
346    not driven by bottom-up processes acting on lower taxa, as Zhang *et al.* (2018b) found.

347        Another formulation of the species-energy hypothesis suggests that highly productive  
348    ecosystems contain abundant resources that increase the number of available niches, thereby  
349    allowing species to specialize on a few resource types and leading to increased species co-  
350    existence and richness (Evans *et al.* 2005a, b; Pautasso & Gaston 2005). If this hypothesis holds,  
351    we would expect positive correlations between GPP and raw species richness for all trophic  
352    groups, positive correlations between GPP and the relative contribution of diet specialists to  
353    richness, and negative correlations between GPP and diet generalist relative richness. If we

354 consider trophic grouping as a coarse scale indicator of diet specialization, with predators and  
355 herbivores representing specialists because they feed on only plant or animal tissue and  
356 omnivores as generalists because they feed on both, we indeed observe these trends in our data.  
357 Additionally, at very low GPP, the relative contribution of omnivores to diversity is similar to  
358 that of herbivores, but then declines as GPP increases. However, we recognize substantial  
359 variation in the degree of diet specialization within trophic groups, and more refined diet  
360 specialization analyses are required to test this hypothesis thoroughly.

361 Alternative versions of the species-energy hypothesis suggest that temperature should be  
362 the strongest correlate with diversity because, at higher temperatures, species can switch energy  
363 investments from thermoregulation to reproduction, thereby maintaining larger populations and  
364 decreasing extinction risk (Evans *et al.* 2005b). Our findings do not necessarily refute this  
365 version of the species-energy hypothesis, as we found a positive relationship between mammal  
366 richness and MAT between temperatures of -10 and 20 °C. While GPP was more important than  
367 MAT for explaining total mammal richness, predator richness, and herbivore richness, the  
368 opposite was true for omnivore richness. Furthermore, the relative contribution of omnivores to  
369 diversity generally increased with MAT, suggesting that some temperature-driven effect other  
370 than productivity is important for driving omnivore richness.

371 We did not find support for our hypothesis that omnivore richness is more tolerant to  
372 environmental harshness and heterogeneity than predator and herbivore richness. We considered  
373 harsh environments to be areas with very high or very low MAT (i.e., extreme heat or extreme  
374 cold) and/or low MAP (drought stricken). There was no indication that omnivore richness  
375 benefited in harsh conditions; instead, they appeared less tolerant. Omnivore richness peaked at  
376 milder MAT levels than herbivores and predators, and their relative contribution to diversity was

377 lowest where annual precipitation was minimal. In contrast, predators appear to be reasonably  
378 tolerant of harsh temperature conditions as they maintained high richness levels at MAT  
379 exceeding 25 °C, and the relative contribution of predators to diversity increased at extreme  
380 temperatures. Additionally, predators accounted for 35.4% of the mammals in our data set,  
381 representing 46.75% of per pixel richness on average. The overrepresentation of predators  
382 indicates they are more cosmopolitan than herbivores and omnivores, potentially due to an  
383 ability to tolerate a broader range of environmental conditions.

384 We considered precipitation seasonality, temperature seasonality, and isothermality to  
385 represent temporal environmental heterogeneity. We found that environmental heterogeneity in  
386 the form of precipitation seasonality, an uncommonly explored predictor variable in mammal  
387 diversity studies, was very important for predicting mammal richness as a whole. Precipitation  
388 seasonality emerged as the most important explanatory variable for the richness of all trophic  
389 groups, with maximum richness at intermediate seasonality values. Intermediate levels of  
390 precipitation seasonality correspond to regions that receive precipitation seasonally with a  
391 defined, but often short dry season (Walsh & Lawler 1981; O'Donnell & Ignizio 2012). Total  
392 mammal richness was lowest in regions of low seasonality (i.e., precipitation evenly distributed  
393 throughout the year) and high seasonality (i.e., areas with long dry/short wet seasons). Again, we  
394 found no evidence that omnivore richness disproportionately benefited from more heterogenous  
395 conditions. However, we did find evidence that predators benefit from intermediate levels of  
396 precipitation and that omnivores appeared to be replaced by predators as precipitation seasonality  
397 increased from low to intermediate values.

398 The peaking of mammal richness in regions with seasonal precipitation supports the  
399 theory that intermediate levels of environmental heterogeneity promote species diversity (Adler

400 & Drake 2008; Tonkin *et al.* 2017). Ecological theory and modeling studies have indicated that  
401 temporal fluctuations in environmental conditions promote more species co-existence than stable  
402 environments as long as the oscillations occur with intermediate and predictable frequency (Liu  
403 *et al.* 2021). Seasonal variability satisfies these criteria and likely promotes species co-existence  
404 by creating temporal niches and minimizing fitness differences between species with different  
405 competitive abilities, particularly in predators and herbivores (Chesson 2000; White *et al.* 2010).

406 In contrast, low temporal environmental variability may limit species co-existence by  
407 promoting competitive exclusion of inferior competitors, while high variability limits diversity  
408 by forcing species to endure long periods of suboptimal conditions that could increase extinction  
409 risk (Adler & Drake 2008; Liu *et al.* 2021). There is empirical support for the positive impact of  
410 seasonal fluctuations of various environmental conditions, including precipitation seasonality, on  
411 species diversity among stream invertebrates, waterfowl, and small mammals (Asher & Thomas  
412 1984; Dalby *et al.* 2014), an effect that appears to be driven by seasonal turnover in species  
413 assemblages. Our results indicate that this phenomenon influences mammal diversity globally  
414 and likely affects how diversity is structured among trophic groups by favoring predators at the  
415 expense of omnivores. Under climate change, drought frequency, severity, and spatial extent are  
416 projected to increase in many regions, including areas that currently experience intermediate  
417 precipitation seasonality (Parmesan *et al.* 2022). Such changes will shift these regions to higher  
418 values of the precipitation seasonality index, potentially making these regions more vulnerable to  
419 biodiversity loss.

420 **Conclusions**

421 Past studies on latitudinal gradients in species diversity have largely overlooked the role  
422 of trophic ecology on the distribution of species. Our results show that diversity patterns in

423 mammal trophic groups are not randomly distributed and that different ecological drivers  
424 influence predators, omnivores, and herbivore diversity patterns. Overall, our results suggest that  
425 temporal environmental heterogeneity (particularly in precipitation) and basal energy availability  
426 (e.g., GPP) are important drivers of total mammal richness, the richness within trophic groups,  
427 and the relative contribution of each trophic group to total richness.

428        Although our models accounted for 73-92% of the variability in mammalian richness  
429 across trophic groups, the best-performing models always included latitude, indicating that  
430 environmental heterogeneity and basal energy availability do not fully explain the observed  
431 patterns in trophic diversity. As a result, we cannot discount other hypotheses that attempt to  
432 explain diversity patterns with evolutionary (e.g., cradle hypothesis), anthropogenic, or other  
433 ecological causes. In particular, anthropogenic activities have led to non-random species  
434 extinctions and invasions, and can also modify the environmental variables we found to be  
435 associated with richness patterns (Pacifici *et al.* 2020). Thus, the diversity patterns we identify  
436 here may reflect natural processes and human impacts.

437        The most important drivers of total mammal richness were precipitation seasonality and  
438 GPP, which are both predicted to shift under climate change. Furthermore, we found that areas  
439 experiencing intermediate levels of environmental heterogeneity expressed the highest trophic  
440 richness. However, climate projections and recent weather phenomena indicate that the intensity  
441 and duration of extreme weather events will increase (Diffenbaugh *et al.* 2017), likely resulting  
442 in higher environmental heterogeneity in many systems. Such changes could negatively impact  
443 terrestrial mammal richness and potentially increase the relative contribution of predators to  
444 mammal diversity.

445        Given the strong relationship between precipitation seasonality and terrestrial mammal  
446 richness, our findings indicate the need for additional research on the impacts of precipitation  
447 seasonality on local and global biodiversity. Precipitation seasonality is expected to shift in many  
448 ecosystems under climate change (Pascale *et al.* 2016; Breinl *et al.* 2020). However, projected  
449 changes to precipitation seasonality have been understudied compared to other potential impacts  
450 of climate change (Parmesan *et al.* 2022), making it difficult to predict how trophic richness  
451 could vary geographically under changing precipitation regimes. Therefore, better projections of  
452 precipitation seasonality and concomitant changes to biodiversity are paramount for conservation  
453 and resource management.

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461 for his advice on random forest modeling.

462        **Data Availability**

463        The compiled dataset used for analysis in this study can be found in an online repository at  
464 <http://dx.doi.org/10.17632/pw8r7jh8kw.1>

465

466 **Figure Captions**

467 **Figure 1.** Global geographic patterns of species richness for a) herbivores, b) omnivores, and c)  
468 predators. Richness was assessed at a pixel resolution of  $30 \times 30$  km.

469 **Figure 2.** Distribution of terrestrial mammal species richness by latitude a) and relative  
470 contribution to the richness of each trophic group b). Richness was assessed at a pixel resolution  
471 of  $30 \times 30$  km, and distributions were fit using the GAM function within the ggplot2 package in  
472 the R statistical computing environment.

473 **Figure 3.** Global geographic patterns of species richness overrepresentation for herbivores,  
474 omnivores, and predators. In some regions, two trophic groups are overrepresented. A trophic  
475 group is considered to be overrepresented if its richness within a pixel is greater than the value of  
476 its global upper quartile of richness.

477 **Figure 4.** Partial dependency plots depicting marginal effects of random forest models for total  
478 richness of all mammals (first column), species richness within trophic groups (second column),  
479 and relative contribution of each trophic group to total richness (third column). Only  
480 environmental variables retained in the final models are presented. Temperature seasonality was  
481 only retained in models of relative richness contribution and so is not presented for total mammal  
482 richness or within trophic group richness.

483 **Figure S1.** Bivariate correlations between explanatory variables used in random forest models.  
484 Correlations are based on all data points in the global dataset ( $n=134,491$ ), and presented  
485 coefficients are Spearman's  $\rho$ .

## Tables

**Table 1.** Importance of five predictor variables retained in final random forest models for raw species diversity. Importance is measured as percent increase in mean square error.

	All Mammals	Predators	Omnivores	Herbivores
Precipitation Seasonality	133.7	126.3	132.9	130.8
Latitude Coordinate	99.8	87.6	86.0	98.1
Gross Primary Production	91.7	102.4	85.1	88.1
Mean Annual Temperature	70.1	75.1	109.1	68.8
Mean Annual Precipitation	60.7	72.6	91.9	60.9

**Table 2.** Importance of six predictor variables retained in final random forest models explaining relative contribution (%) of each trophic group to total mammal richness. Importance is measured as percent increase in mean square error.

	Predators	Omnivores	Herbivores
Gross Primary Production	97.3	50.7	74.8
Precipitation Seasonality	81.4	57.0	56.4
Temperature Seasonality	78.5	27.4	64.1
Mean Annual Temperature	64.6	41.7	36.8
Mean Annual Precipitation	63.1	52.3	62.7
Latitude Coordinate	66.9	33.7	62.3

Figure 1

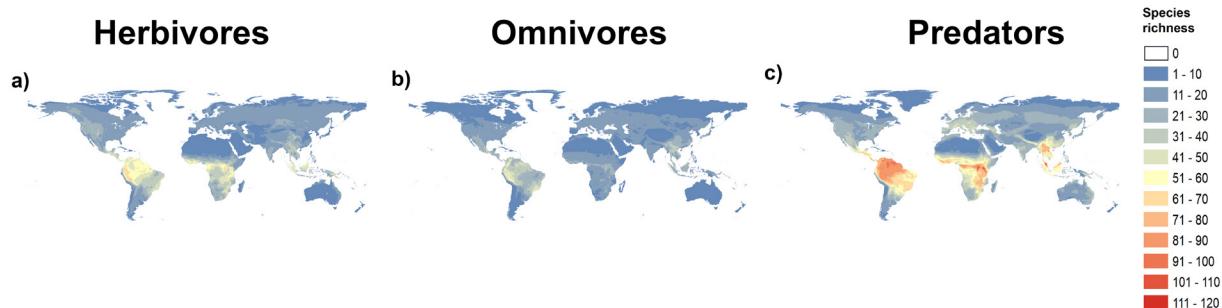


Figure 2

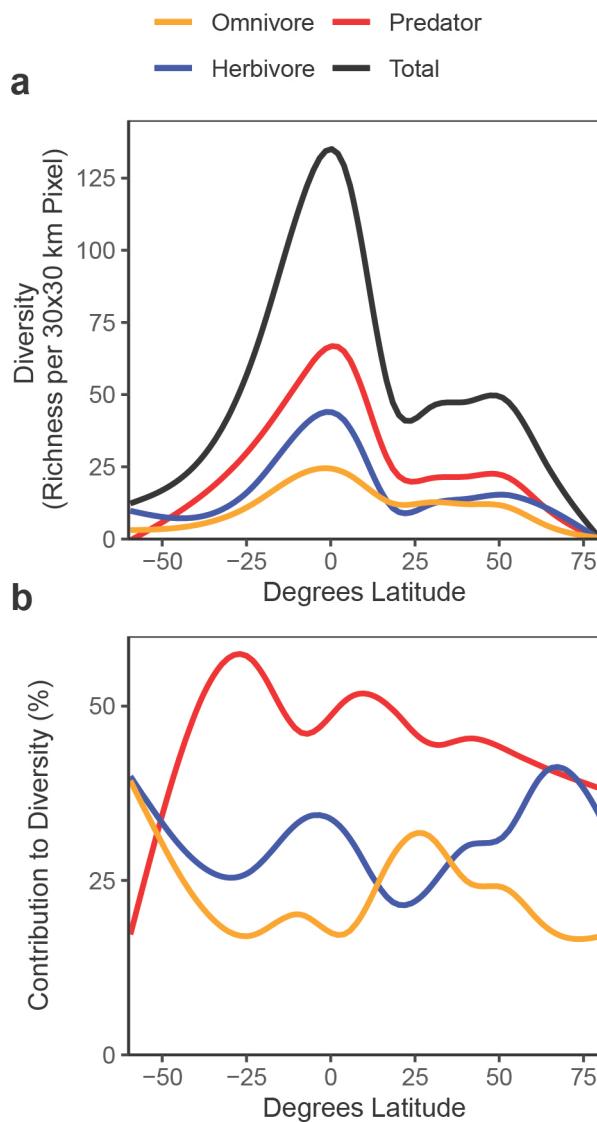


Figure 3

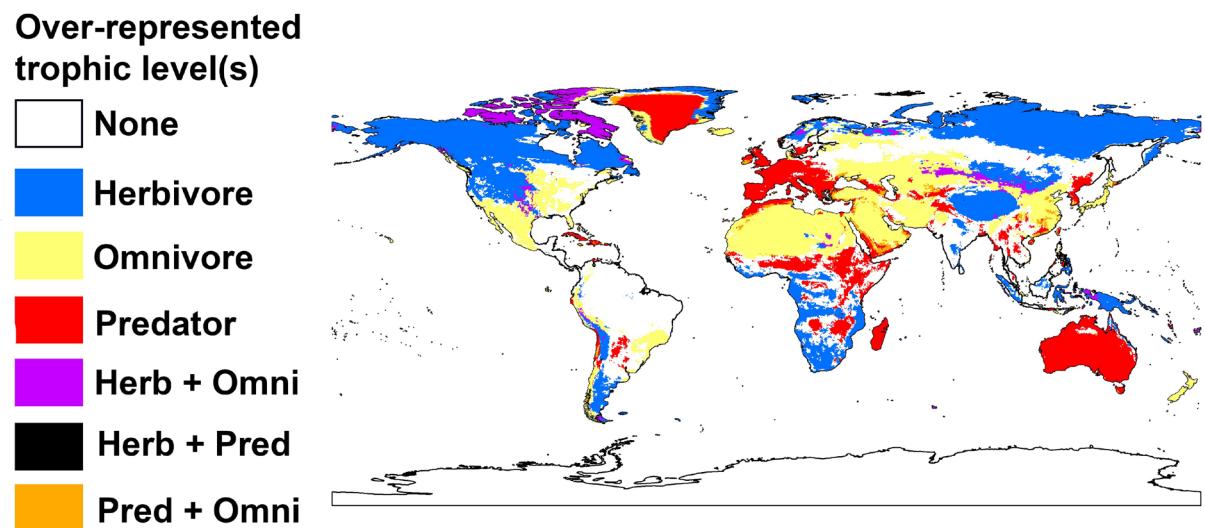
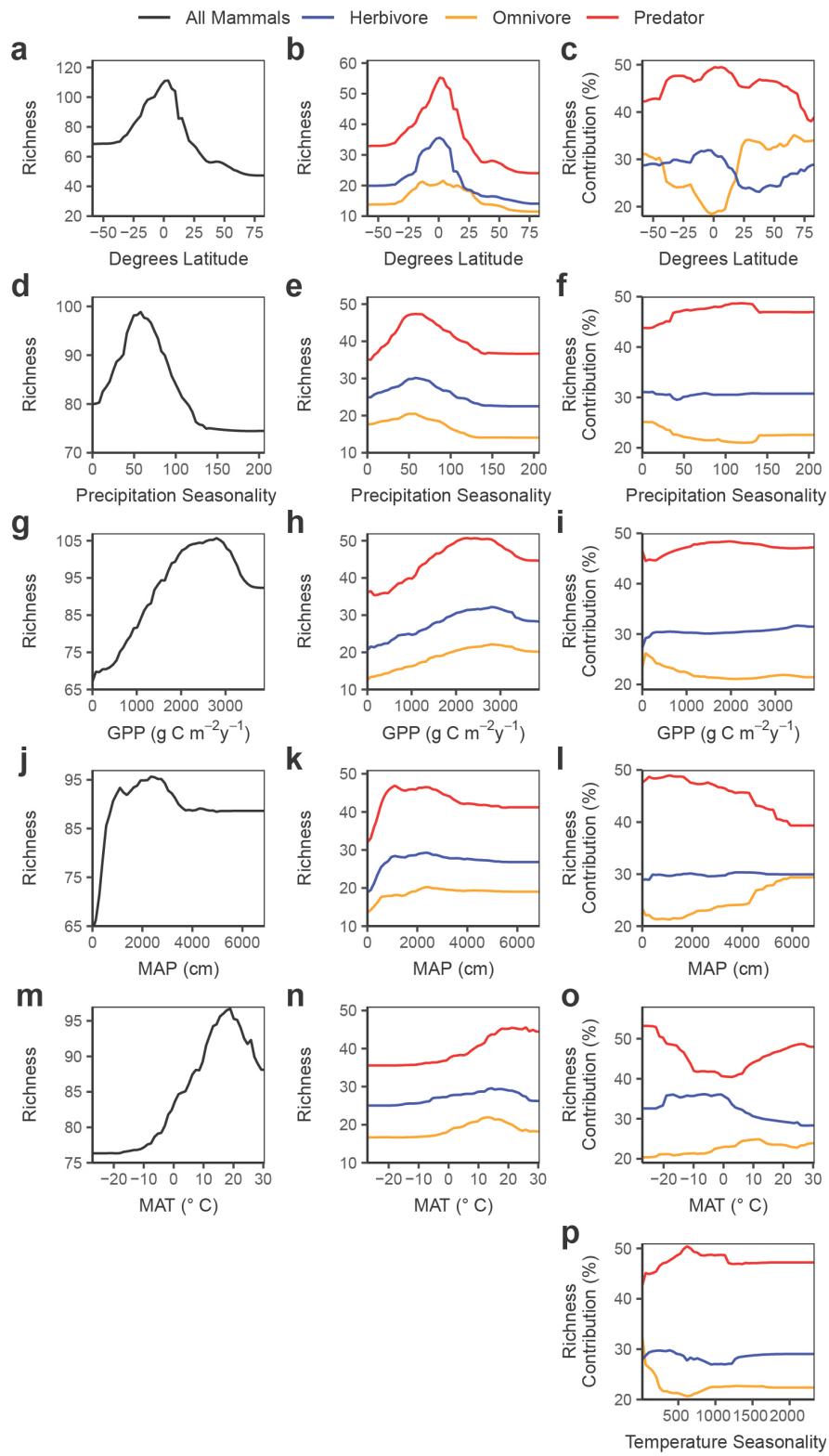


Figure 4



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