1 Does multiple nutrient enrichment impact the stability of grassland biomass

2 production?

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76 ABSTRACT

77 Nutrient enrichment can simultaneously increase and destabilize plant biomass production, with 78 co-limitation by multiple nutrients potentially influencing these effects. Here, we test how 79 factorial additions of nitrogen (N), phosphorus (P), and potassium with essential nutrients (K+) 80 affect the stability (mean/standard deviation) of aboveground biomass in 34 grasslands over 81 seven years. Destabilization with fertilization was prevalent and was largely driven by single 82 nutrients, rather than synergistic nutrient interactions. N-based treatments increased mean 83 production by 21-51% but increased temporal variability by 40-68% and so consistently reduced 84 stability. P alone increased variability and instability without altering mean biomass, while K+ had 85 no general effects. Declines in stability were largest in the most nutrient-limited grasslands, or 86 where nutrients reduced species richness or intensified species synchrony. We show that adding 87 limiting nutrients can differentially impact mean-variability proportionality and that N and P in 88 particular magnify fluctuations in inter-annual productivity, even if biomass continues to 89 increase.

91 INTRODUCTION

92 Biomass production is a critical ecosystem function that supports services such as carbon storage 93 and food supply (Kremen 2005; Cardinale et al. 2012; Gounand et al. 2020). Consequently, 94 understanding how global changes impact the rate and stability of ecosystem biomass 95 production – i.e. the ratio of the temporal mean of biomass production to its temporal standard deviation ($S = \mu/\sigma$; Tilman 1999; Tilman *et al.* 2006) - is an important undertaking (Koerner *et al.* 96 97 2016; Avolio et al. 2020). Terrestrial biomass may be limited by the availability of a single resource 98 like water (Huxman et al. 2004; Morgan et al. 2016), nitrogen (N; LeBauer & Treseder 2008), or 99 phosphorus (P; Hou et al. 2020), or co-limited by multiple resources that interactively affect 100 production (Elser et al. 2007; Harpole et al. 2011; Fay et al. 2015; Kaspari & Powers 2016). 101 Accordingly, pervasive human-driven alterations of nutrient availability can strongly affect rates 102 of biomass production and associated ecosystem services (Tilman 2001; Tilman & Lehman 2001). 103 Nutrient enrichment can also destabilize biomass production over time by altering community attributes that contribute to its stability, such as species richness and synchrony (Hautier et al. 104 105 2014, 2020; McCann et al. 2021). However, while the co-limitation of biomass production by 106 multiple nutrients is well demonstrated, we lack understanding of the independent and 107 interactive effects of different nutrients on its temporal stability.

Processes operating among individuals, populations, and communities interact to regulate ecosystem biomass production and its response to nutrient enrichment. For individual plants, nutrient availability regulates biomass production by imposing physiological constraints on growth (Droop 1974). Subsequently, guided by species niche and fitness differences and species richness, interactions among individuals and species determine how the effects of 113 nutrients on individual growth add up to community level biomass production (Mellinger & 114 McNaughton 1975; Bakelaar & Odum 1978; Loreau 2010; Carroll et al. 2011). Resource limitation 115 influences this dynamic by providing opportunities for niche differences between species to 116 contribute to species coexistence (Levin 1970; Chesson 2000; Harpole & Tilman 2006). As such, 117 multiple nutrient enrichment can reduce species richness by constraining niche dimensionality 118 (Harpole & Tilman 2007; Harpole et al. 2016, 2017) and causing the competitive exclusion of 119 species that are less competitive in the new resource environment (Braakhekke & Hooftman 120 1999; Danger et al. 2008). This dynamic drives changes in community biomass production that 121 depend on the number and identity of enriching nutrients (Harpole et al. 2011; Fay et al. 2015).

122 The temporal variability of community biomass is driven by processes that control species 123 responses to environmental fluctuations (Tilman 1996; Loreau 2010). Declines in species richness 124 generally reduce the stability of biomass production (Hautier et al. 2015) by increasing the chance 125 of losing those species that can maintain function in a given environmental context (Loreau 126 2010). For example, temporal instability is frequently exacerbated when species that tolerate 127 extreme conditions such as drought or fire are lost (MacDougall et al. 2013). Community stability 128 is also influenced by changes in species synchrony, which regulate how species-level variability 129 scales to community-level variability to produce functional compensation (Hector et al. 2010; 130 Hautier et al. 2014; Gilbert et al. 2020; Valencia et al. 2020a). Synchrony measures the covariance 131 of species' responses to a fluctuating environment and is driven by niche differences and life-132 history processes that impact the capacity for response (Loreau 2010). Together, these processes 133 guide fluctuations in species competitive rankings and result in compensatory dynamics, where 134 a functional decline by one species reduces competitive suppression of others, allowing them to

increase their biomass production (Brown *et al.* 2016). Consequently, nutrient enrichment and
 multiple-nutrient interactions can influence the temporal variability of community biomass
 production by influencing the number of niche differences that maintain species diversity and
 guide species synchrony.

139 As the mean and temporal variability of biomass production may often be driven by these 140 mechanisms of species complementarity and species synchrony, respectively, there is the 141 potential for them to show different magnitudes of response to nutrient enrichment (lves et al. 142 2000; Cardinale et al. 2013; Kohli et al. 2019). Transitions in the relationship between the mean 143 and variability of biomass production over time define changes in S and are an important 144 component of ecological stability (Ives & Carpenter 2007; Donohue et al. 2013). Declines in S 145 indicate increases in proportional variability and correspond to less consistent ecosystem 146 function (Donohue et al. 2016) and an increased likelihood of crossing minimum or maximum 147 thresholds (Carnus et al. 2014) where irreversible change in community composition or functioning can occur (Beisner et al. 2003; Carpenter & Brock 2006). This transpires in enriched 148 149 systems if, despite higher average rates of production, concurrent increases in temporal 150 fluctuations enhance the risk that biomass production falls below a critical threshold (Scheffer et 151 al. 2001; Beisner et al. 2003; Carnus et al. 2014). While this paradox of enrichment dynamic is 152 well described in the theoretical literature (Rosenzweig 1971), its interaction with co-limitation 153 in empirical systems remains poorly characterized.

Here, we assessed the effects of individual and combined N, P and Potassium + essential nutrients (K+) enrichment of 34 grassland sites spanning six continents over seven years. We hypothesized that the effect of chronic nutrient enrichment on the temporal stability of biomass 157 production depends on nutrient identity and multiple-nutrient interactions and is partly driven 158 by factors including underlying co-limitation, local precipitation, species loss, and species 159 synchrony. We evaluated our hypotheses in three stages. First, we assessed temporal stability under each treatment to determine if different individual and combined nutrient inputs had 160 161 different stability effects. Second, we assessed the mean and variability of biomass production 162 under each treatment to determine if changes in stability were mean- or variability-driven 163 (Carnus et al. 2014; Kohli et al. 2019). We expected to frequently observe variability-driven 164 destabilization because the mechanisms that control variability have shown strong responses to 165 enrichment (Hautier et al. 2014; Gilbert et al. 2020). Alternatively, synergistic responses to 166 multiple nutrient inputs could cause large increases in mean biomass that would mitigate against 167 destabilization (Harpole et al. 2011; Kaspari & Powers 2016). Finally, we assessed site differences 168 in stability responses and tested three potential mechanisms of different destabilization effects 169 among sites. First, we examined whether destabilization would be stronger where nutrient 170 limitation was stronger. Consequently, we expected N and P additions would have larger 171 destabilization effects than K+ because they are more commonly limiting (LeBauer & Treseder 172 2008; Fay et al. 2015; Hou et al. 2020). Similarly, we expected multiple nutrient enrichment would 173 cause greater destabilization than single nutrient inputs given frequent co-limitation (Harpole et 174 al. 2011; Fay et al. 2015). Second, we examined whether enrichment of sites with lower average 175 annual precipitation would have less effect on temporal variability because biomass is likely to 176 be more limited by water availability than nutrients (Paruelo et al. 1999; Huxman et al. 2004; 177 Morgan et al. 2016). Finally, we tested if changes in stability are driven by changes in species 178 richness (Hautier et al. 2015) or species synchrony (Loreau 2010; Valencia et al. 2020a), such that stability decreases where species are lost or synchrony increases. We predicted that multiple nutrient treatments would be more destabilizing than single nutrient treatments because they are likely to cause greater species loss (Harpole & Tilman 2007; Harpole *et al.* 2016) and impose greater constraints on the niche differences that permit asynchrony.

183

184 METHODS

185 Experimental design and nutrient additions

We assessed the mean, interannual variability and stability of aboveground biomass production in 34 grassland sites that received standardized annual inputs of N, P and K+ for 7 years (Table S1). These data are from the globally distributed Nutrient Network (NutNet) experiment, described in Borer et al (2014). Our decision to study a seven-year treatment period was made to balance duration, thereby allowing temporal dynamics to unfold, with replication of sites, thereby analyzing a broad range of grassland ecosystems. We tested the sensitivity of our analyses to all subsets between one-three and one-twelve years (Table S2).

193 The majority of sites contain 3 blocks (but ranges from 1 to 6; Table S1) comprising 5 x 5 194 m treatment plots that receive one of the seven possible factorial combinations of N, P and K+ 195 fertilizer, in addition to an unfertilized control plot. Plots are arranged in a randomized block 196 design. Nutrients were applied at a standardized yearly rate, using 10 g N (time-release urea) m⁻ ² yr⁻¹, 10 g P (triple-super phosphate, which also includes Ca) m⁻² yr⁻¹, and 10 g K (potassium 197 sulphate, which also includes S) m⁻² yr⁻¹. In the first year only, all K plots also received 100 g m⁻² 198 199 of a macro- and micronutrient mix, comprising: 15% Fe, 14% S, 1.5% Mg, 2.5% Mn, 1% Cu, 1% Zn, 200 0.2% B, and 0.05% Mo – forming the K+ treatment. All plots included in this analysis were open to herbivory. Site-level mean annual precipitation (MAP) and MAP variability data were obtained
from the WorldClim Global Climate database (Version 1.4; Hijmans *et al.* 2005).

203

204 Metrics

We harvested aboveground biomass annually from two 1 m x 10 cm strips in each plot at the sitespecific time of peak biomass production. Live biomass was separated from dead biomass, dried at 60°C and weighed to provide estimated biomass production in g m⁻² year⁻¹ for each plot (Borer *et al.* 2014). Harvest strips were relocated within a plot each year to avoid a clipping effect. We calculated the temporal mean (μ) of biomass for each plot as the mean of annual biomass measurements.

211 Ecosystems responding to chronic nutrient inputs can display directional trends in 212 biomass (Seabloom et al. 2021) that may influence stability metrics (Tilman et al. 2006; Lepš et 213 al. 2019). To focus our analysis on interannual fluctuations in biomass production, we detrended 214 our data by taking the residuals from a linear regression of biomass over years of treatment in each plot (Tilman *et al.* 2006). Subsequently, the standard deviation of residuals in a plot (σ_{detr}) 215 216 was used as a measure of interannual variability and in calculating a single detrended S value for each plot, as: $S_d = \mu / \sigma_{detr}$. Values of S_d are presented as percentages in figures to indicate 217 218 proportional interannual variability (relative to treatment plot μ).

Plant species richness and percent cover to 1% were surveyed in permanent 1 m² quadrats in each plot, based on visual assessment. Surveys were conducted annually or biannually according to the growing season period at each site. We used these data to calculate the temporal mean of species richness for each plot. Additionally, we used data for species' 223 percent cover to calculate detrended measures of temporal species synchrony in each plot, using 224 the calc_sync function (Lepš et al. 2019) in R (v 3.6.3; R Foundation for Statistical Computing). 225 This function detrends the ϕ synchrony metric (Loreau & de Mazancourt 2008) by aggregating ϕ 226 values from a moving 3-year window (φ_{t3} ; Lepš *et al.* 2019). This mitigates the influence of 227 directional trends, which can cause correlations in species abundances over time that are 228 separate from the year-to-year fluctuations that drive synchrony (Lepš et al. 2019; Valencia et al. 229 2020b). Values of ϕ_{t3} are bounded between 0 and 1, where 0 indicates perfect asynchrony and 1 230 indicates perfect synchrony.

231 We calculated treatment effects on each of these metrics using natural-log response 232 ratios, as: $LRR = \ln(treatment/control)$. This effect size highlights the change in the value of 233 a metric in a treatment plot relative to its value in a control plot located within the same 234 experimental block. It standardizes treatment effects across sites with different initial rates of biomass production or variability. In addition, using LRRs centered values around zero and helped 235 236 data meet the conditions of normality that were assumed for our analysis (Hedges et al. 1999). 237 We back-transformed LRR estimates and converted them to % change relative to the control 238 treatment for presentation in figures.

239

240 Analysis

We used linear mixed effects models to test how different nutrient inputs impacted the S_d , μ , and σ_{detr} of grassland biomass. We assessed each metric using the model:

243
$$LRR_{ijk} = \beta_i + b_j + \varepsilon_{ijk}$$
[1]

244 These models provide estimates for the fixed effect at each of the 7 levels of nutrient treatment 245 (i), estimates of random between-site variation in the response (j), and estimates of the within-246 site variation (ijk) that results from observation of multiple blocks at each site. We chose this 247 model structure after trialing more complicated models including random slopes for the effect 248 of treatment within each, but these models did not converge. We also used model 1 to test for 249 nutrient effects on the LRRs of species richness and species synchrony. For each response, we 250 tested interactions between nutrients by replacing the nutrient term in model 1 with all two- and 251 three-way interactions between dummy coded N, P and K+ factors.

252 We assessed the dependence of stability responses on mean biomass responses using the 253 model:

$$LRR. S_{d,ijk} = \beta_{1,i} + b_j + \beta_{2,i}\chi_i + \varepsilon_{ijk}$$
^[2]

255 Where χ_i is the response of biomass in treatment *i* at site *j*. We built these models and obtained 256 estimates of model parameters by maximum likelihood estimation using the lme4 package (Bates 257 *et al.* 2015) in R.

258 We used a combination of methods to evaluate both the statistical and biological 259 significance of our effect size estimates (Nakagawa & Cuthill 2007). To assess statistical 260 significance, we present mean effects with their associated 95% confidence intervals and provide 261 P-values that indicate whether a treatment effect was significantly different from zero (at alpha 262 < 0.05). To interpret biological significance, we compare our results with effect sizes observed in 263 previous studies of grassland biomass production. For example, we used stability effects 264 observed in a meta-analysis of experimental biodiversity studies (Hautier et al. 2015) as a 265 benchmark for interpreting nutrient effects on stability in our analysis (Table S3). We used this approach to determine the number of sites with predicted stability, mean biomass, and standard
deviation responses that passed thresholds of biological significance.

268 To test for potential drivers of among-site variation in nutrient treatment effects, we built 269 a second model for each of the three biomass responses. Fixed effect predictors included main 270 effects for plot-level LRRs of plant species richness and species synchrony, site-level MAP and 271 MAP variability, nutrient treatment, and the interaction effect of nutrient treatment with each 272 additional predictor. We evaluated the contribution of these factors to changes in the mean, 273 variability and stability of biomass production using a model selection approach. We used the 274 dredge function in the MuMIn package (Bartoń 2009) to determine which set of predictors and 275 interactions best explained the responses. We selected all predictors included in models within 276 four AIC of the most parsimonious model and obtained their full averages and associated 277 summary statistics using the model.avg function (Bartoń 2009).

278

279 **RESULTS**

Do nutrient identity and nutrient interactions impact the temporal stability of grassland
 biomass production?

At the global scale, seven years of single nutrient N enrichment decreased S_d by 14% relative to control plots (LRR_N = -0.16 P < 0.001; Fig. 1a, Table S4). Inputs of P were also generally destabilizing (LRR_P = -0.12, P = 0.011). In contrast, K+ did not significantly affect S_d (LRR_{K+} = -0.04, P = 0.322).

286 Multiple nutrient input treatments destabilized biomass (Fig. 1a, Table S4) by 13% with 287 NP (LRR_{NP} = -0.14, P = 0.003), 15% with NK+ (LRR_{NK+} = -0.17, P < 0.001), and 12% with NPK+ 288 (LRR_{NPK+} = -0.13, *P* = 0.004). However, we did not observe destabilizing interactions between any 289 nutrients (Table S5). Instead, stability in NP, NK+ and NPK+ plots was equivalent to the stability 290 of single-nutrient N and P input plots. Further, sub-additive interactions between N and P acted 291 to stabilize biomass relative to the potential additive effect of N and P inputs (LRR_{N*P} = 0.14, *P* = 292 0.045).

293

294 Do nutrient identity and nutrient interactions impact the temporal mean and variability of 295 grassland biomass production?

At the global scale, increases in mean biomass following nutrient inputs were outpaced by greater increases in temporal variability (Fig. 1b-c), a trend that defines decreases in S_d . Inputs of N increased mean biomass production by 21% (LRR_N = 0.19, P < 0.001), while P (LRR_P = 0.06, P= 0.149) and K+ (LRR_{K+} = -0.001, P = 0.926) did not significantly impact mean biomass globally (Fig. 1b, Table S6). Concurrently, N additions increased the interannual variability of biomass by 40% (LRR_N = 0.36, P < 0.001) and P additions by 19% (LRR_P = 0.17, P = 0.007; Fig. 1c, Table S7). K+ inputs had no significant effect on σ_{detr} (LRR_{K+} = 0.04, P = 0.5).

Multiple nutrient addition effects on the mean and variability of biomass production were larger than single nutrient effects. Inputs of NP, NK+, and PK+ increased mean biomass by 48% (LRR_{NP} = 0.39, P < 0.001), 28% (LRR_{NK+} = 0.26, P < 0.001) and 16% (LRR_{PK+} = 0.15, P < 0.001), respectively (Fig. 1b, Table S6). Concurrently, they increased interannual variability by 68% (LRR_{NP} = 0.52, P < 0.001), 49% (LRR_{NK+} = 0.42, P < 0.001) and 21% (LRR_{PK+} = 0.2, P = 0.002), respectively (Fig. 1c, Table S7). Simultaneous NPK+ additions increased mean biomass by 51% (LRR_{NPK+} = 0.41, 309 P < 0.001), exceeding the effect of all one- or two-nutrient treatments. Concurrently, NPK+ 310 increased interannual variability by 68% (LRR_{NPK+} = 0.53, P < 0.001).

We observed a general synergistic co-limitation of mean biomass production by N and P (LRR_{N*P} = 0.14, P = 0.011, Table S5). There was also a notable interaction between P and K+ (LRR_{P*K+} = 0.1, P = 0.071) that increased biomass in PK+ plots relative to controls. In contrast, we did not observe any significant nutrient interactions for temporal variability (Table S5). Instead, multiple nutrient effects on interannual variability were driven by significant effects of N and P that produced additive increases in σ_{detr} (Table S5).

317

318 How consistent were effects among the 34 globally distributed grassland sites?

319 Destabilization most frequently occurred in response to treatments containing N. Inputs of N 320 alone decreased stability by more than 10% relative to control plots at 23 sites and by more than 321 20% at 8 sites (Table S8). These eight sites were globally distributed and did not follow a 322 detectable geographical trend. Similarly, large biomass and variability responses were most 323 frequently observed in response to N treatments. Mean biomass increased by over 25% with 324 NPK+ at 31 sites and by over 50% at 18 sites (Table S8). Variability increased by over 25% at 33 325 sites and by over 50% at 26 sites. Multiple nutrient treatments produced mean biomass and 326 variability responses more frequently than inputs of single nutrients (Table S8).

327

328 Does nutrient limitation status impact the destabilization potential of nutrient enrichment?

329 Across the study, sites with larger biomass responses, indicative of stronger underlying nutrient

330 limitations, were associated with larger destabilization responses to nutrient enrichment (Fig. 2,

Table S9). This relationship held for all nutrient combinations except NP and NPK+ (Table S9).

332

333 Does average site precipitation influence the response of temporal stability to nutrient 334 enrichment?

Average site precipitation did not strongly drive stability effects; MAP and MAP variability were retained in the model selection procedure for the stability, mean biomass, and temporal variability models but were not significant predictors of these responses (Table S10).

338

339 Do nutrient effects on species richness or species synchrony influence the response of temporal
 340 stability to nutrient enrichment?

Species richness declined with N and all multiple nutrient treatments but was not generally impacted by P or K+ alone (Fig. 3a, Table S11). Species loss was highest with simultaneous NPK+ inputs, which reduced species richness by 16% (LRR_{NPK+} = -0.17, P < 0.001) across the study. Species synchrony did not display a clear directional response to any nutrient treatment at the global scale (Fig. 3b, Table S12).

 S_d was positively associated with changes in species richness (*P* = 0.009; Fig. 4a, Table S10). Specifically, species loss was associated with decreased stability of biomass production whilst species gains were associated with increased stability. In contrast, overall mean biomass responses were negatively associated with species richness such that biomass was most likely to increase where species were lost (*P* = 0.001; Fig. 4b, Table S10). Similarly, change in the interannual variability of biomass production was negatively associated with species richness such that plots displayed greater increases in temporal variability where species loss was higher (P < 0.001; Fig. 4c, Table S10).

There was a negative relationship between species synchrony and S_d responses (P = 0.052; Fig. 5, Table S10). Consequently, destabilization of community biomass was more likely where nutrients caused species synchrony to increase, and stabilization more likely where nutrients promoted asynchrony (Fig. 5). Species synchrony was retained in the model set for the mean and variability of biomass but was not a statistically significant predictor of these responses (Table S10).

360

361 **DISCUSSION**

362 Do nutrient identity and nutrient interactions impact the temporal stability, mean and 363 variability of grassland biomass production?

364 Grassland biomass production is often limited or co-limited by nutrients (Harpole et al. 2011; Fay 365 et al. 2015). However, understanding of how co-limitation dynamics affect grassland stability is 366 still lacking. This is crucial to assessing the reliability of critical ecosystem services or threat of 367 eventual ecological collapse in nutrient enriched environments (McCann et al. 2020). Our 368 analyses showed that N and multiple nutrient enrichment generally increased mean biomass 369 production but destabilized it by inducing even greater increases in temporal variability. This 370 signals some years of elevated biomass production but also that the magnitude of fluctuations 371 between years increased with N supply. Enrichment of P did not generally increase mean biomass 372 but still increased its temporal variability and therefore had a general destabilizing effect. In 373 contrast, K+ enrichment did not change either the mean or variability of biomass, and therefore374 did not impact stability at the global scale.

375 Multiple nutrient treatments revealed interactive effects on mean biomass production, 376 including a general synergistic interaction between N and P, consistent with previous 377 observations of co-limitation (Harpole et al. 2011; Fay et al. 2015; Kaspari et al. 2017). However, 378 this did not directly translate to stability responses. Multiple nutrient inputs did not drive 379 transitions in mean-variability scaling beyond the effects of single N or P inputs. Consequently, 380 nutrient interaction effects on stability were generally sub-additive (i.e. less than the sum of 381 multiple single-nutrient effects; Harpole et al. 2011). This mitigated against compounding 382 destabilization effects that could occur if all added nutrients exerted additive or synergistic 383 effects (Harpole et al. 2011).

384 Accordingly, we did not find evidence that multiple-nutrient enrichment was more 385 destabilizing than single-nutrient enrichment at the global scale. This suggests that nutrient input 386 identity and interactions can have consequences for trade-offs between increased grassland biomass production and the risk of destabilization. That is, declining S_d may indicate an increased 387 388 risk of reaching thresholds corresponding to a minimum acceptable value of an ecosystem 389 function (Carnus et al. 2014) where , for example, substantial changes to community composition 390 become likely (Scheffer et al. 2001; Beisner et al. 2003; Carpenter & Brock 2006; McCann et al. 391 2021) or a food production system fails to reach a profit. Our study suggests that, relative to 392 control plots, single N inputs generally produced higher mean biomass production (more reward) 393 but decreased stability (more risk). In contrast, relative to N, multiple-nutrient NP and NPK+ inputs generally produced higher mean biomass but with an equivalent S_d . 394

395 All inputs containing N generally increased temporal fluctuations in biomass production, 396 consistent with previous empirical observation (Hautier et al. 2014) and theoretical work on the 397 paradox of enrichment (Rosenzweig 1971). However, while NP and NPK+ produced the same risk 398 as N-only inputs, they offered greater biomass production; potentially equating to greater 399 delivery of ecosystem services such as food production, food system profitability, biofuels and 400 soil carbon sequestration (Kremen 2005; Gounand *et al.* 2020). Nonetheless, situations where S_d 401 is equivalent but means are different indicate that total interannual variability is greater in the 402 high means group. Variability can be of interest itself as it determines the absolute size of 403 fluctuations in ecosystem services (Kohli et al. 2019). In this study, absolute variability was higher in NP and NPK+ treatments than with N only, despite equivalent S_d values; a result that could 404 405 still translate to altered nutrient cycling or secondary production (Kohli et al. 2019). These 406 intricacies highlight the necessity of being clear about the aspects of variability and stability that 407 are of concern in a given context (Carnus et al. 2014; Kohli et al. 2019).

408 Here, explicit consideration of nutrient effects on both the mean and variability of 409 biomass indicated a general effect in which adding limiting nutrients increased mean biomass 410 production but drove disproportionate increases in variability, resulting in variability-driven 411 destabilization. It also revealed that the type of multiple nutrient limitation of the mean 412 (synergistic vs additive; Harpole et al. 2011) was not the same as multiple limitation of temporal 413 variability. While consideration of both the mean and variability components of stability remain 414 relatively uncommon (Carnus et al. 2014; Hautier et al. 2015; Kohli et al. 2019; Avolio et al. 2020), 415 global change drivers do not always appear to alter mean-variability proportionality. For 416 example, Kohli et al. (2019) observed that disturbance of consumer food webs had no effect on

417 the stability of grassland biomass production because the mean and variability responded 418 proportionately. These results show that future work should routinely consider how both the 419 mean and variability of ecosystem processes respond to key global changes (Avolio *et al.* 2020). 420

421 How consistent was nutrient-driven destabilization among **34** globally distributed grassland 422 sites?

423 The magnitude of nutrient-driven changes in stability varied among sites. Twenty sites displayed 424 10% declines in stability following NPK+ enrichment and eight displayed reductions in stability of 425 more than 20%. This effect size has a similar magnitude to previously reported stability 426 responses. For example, declining species richness is a widely acknowledged driver of significant 427 destabilization (Tilman et al. 2006; Isbell et al. 2013; Hautier et al. 2015) and, in a recent meta-428 analysis, experimentally reducing richness from 16 to 2 was also shown to destabilize biomass 429 production by 20% (Hautier et al. 2015). This suggests that the effects of nutrient enrichment on 430 stability can match other key global change drivers. Overall, our study suggests that there is a 431 wide range in the magnitude of grassland stability responses to elevated nutrient supply, but that 432 increasing eutrophication will likely drive more grasslands towards critical thresholds of 433 functioning.

Nutrient effects on mean biomass production were more prevalent across all the sites. At
31 sites, NPK+ enrichment increased mean biomass production by at least 25%. Nutrient
limitation has not appeared this prevalent in previous, shorter assessments. For example,
previous analyses of NutNet data showed that over 3-5 years 12 of 42 sites were not significantly
limited by any combination of N, P and K+ (Fay *et al.* 2015). We included nine of those sites in the

present study and, after seven years, NPK+ had driven >30% increases in biomass production at eight of them. This is consistent with increasing nutrient effects observed over ten years (Seabloom *et al.* 2021) and might be explained by a progressive shift in community composition, as opposed to immediate and possibly transient changes in relative abundance and individual biomass production. Nutrient effects on temporal variability were even more prevalent, with 33 sites increasing variability by 25% and 26 sites increasing variability by 50% following NPK+ enrichment.

446

447 Was there evidence for mechanisms associated with destabilization following nutrient 448 enrichment?

449 Understanding of the mechanisms that determine the potential for different nutrients to 450 destabilize grassland production is crucial to mitigating the risk of destabilization in a given 451 environmental context. In our analysis, there was no clear association of nutrient-driven destabilization with MAP or MAP variability. This contrasts our prediction that site precipitation 452 453 would mediate nutrient effects on biomass stability, based on foundational work suggesting the 454 effect of nutrient limitation on grassland biomass production decreases with increasing moisture 455 limitation (Paruelo et al. 1999; Huxman et al. 2004). However, our finding is consistent with 456 broader observations that nutrient enrichment can destabilize biomass production in both wet 457 (Tilman et al. 2006) and dry (Wang et al. 2017) grassland systems. One explanation for this 458 pattern is that different mechanisms mediate the relationship between moisture availability and 459 stability in wet and dry systems, but each still promote destabilization (Wang et al. 2017). 460 Nutrient enrichment of wet grasslands often reduces stability by reducing species richness and

461 negating positive diversity-stability effects (Tilman *et al.* 2006; Wang *et al.* 2017). In contrast, in 462 dry grasslands, nutrient enrichment may increase the responsiveness of biomass production to 463 occasional wet years, therefore generating larger temporal fluctuations through occasional 464 increases in growth and, ultimately, reducing stability (Wang *et al.* 2017). Studies that assess wet 465 and dry grasslands separately have provided better insight the role of precipitation and moisture 466 availability in mediating nutrient effects on the stability of grassland biomass production (Bharath 467 *et al.* 2020).

468 Our study suggests that destabilization following nutrient enrichment is partly driven by 469 the extent of underlying nutrient limitation, meeting our expectation that destabilization 470 responses would be stronger where nutrient limitation is stronger. One explanation for this is 471 that nutrient limitations may impose constraints on community composition by mediating 472 competition among species (Tilman 1982; Braakhekke & Hooftman 1999). Enrichment of strongly 473 nutrient-limited communities may cause shifts in the outcomes of competitive interactions (Tilman 1982) and changes in community composition that impact community biomass 474 475 production. Despite the overall relationship mean and stability effects, the response of mean 476 biomass was not a significant predictor of stability within the NP and NPK+ treatments, suggesting 477 these inputs also reduced stability where they did not limit biomass production. One explanation 478 for this is that NP and NPK+ enrichment can drive species loss even where those nutrients are 479 not limiting factors (Harpole et al. 2016), thereby causing destabilization through lost diversity-480 stability effects, as discussed below.

481 Our results also supported the prediction that nutrient effects on species richness would 482 contribute to changes in mean biomass production and its stability (Hautier *et al.* 2015). Plots

483 that lost more species typically became more productive but less stable due to proportionally 484 larger increases in the temporal standard deviation of biomass production. This has not been observed in previous analyses of NutNet data (Hautier et al. 2014, 2020). Our current findings 485 486 better match theoretical predictions about diversity-stability relationships, and our ability to 487 detect this here is likely due to the increased power afforded by including observations from all 488 treatment plots (more than just NPK+ plots) and the accumulation of more sites with longer-term 489 data. In addition to species richness effects, we also found that changes in species synchrony 490 following enrichment contributed to changes in stability. Overall, stability was reduced where 491 synchrony increased, in agreement with previous work (Loreau & de Mazancourt 2008; Hautier 492 et al. 2014, 2020; Muraina et al. 2021). However, in contrast to richness, there were no general 493 effects of nutrient treatments on synchrony, nor differences between treatments.

494 In addition to decreased stability in response to nutrient enrichment observed here, our 495 results also suggest a heightened destabilization risk over longer periods of chronic enrichment. 496 Coupled with evidence that species loss can continue for more than a decade with simultaneous 497 NPK+ enrichment (Seabloom et al. 2021), our observation that species richness loss contributed 498 to nutrient-driven destabilization suggests that increased variability could be exacerbated over 499 longer periods. Further, our observation of increased overall mean biomass production with 500 species loss was characteristic of a shift in community composition towards highly productive 501 species (Tilman 1982; Harpole & Tilman 2007; Hautier et al. 2009). Productive species can 502 mitigate against destabilization by maintaining a high temporal mean; however, sites with higher 503 species loss may be more susceptible to future crashes in productivity driven by increased 504 sensitivity to other environmental variability such as drought, herbivory, or fire (MacDougall et al. 2013). This would negate mean-driven stabilization effects and exacerbate destabilization.
This dynamic can lead to a collapse in biomass production over longer timeframes (Isbell *et al.*2013). This long-term effect is more likely with increasing numbers of added nutrients, which can
cause increasing species loss (Harpole & Tilman 2007; Harpole *et al.* 2016). Our findings suggest
this risk is particularly enhanced with N and the NP interaction that drove the greatest species
loss.

511

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522 **REFERENCES**

- 523
- Avolio, M.L., Wilcox, K.R., Komatsu, K.J., Lemoine, N., Bowman, W.D., Collins, S.L., *et al.* (2020).
 Temporal variability in production is not consistently affected by global change drivers
 across herbaceous-dominated ecosystems. *Oecologia*.
- 527 Bakelaar, R.G. & Odum, E.P. (1978). Community and Population Level Responses to Fertilization 528 in an Old-Field Ecosystem. *Ecology*, 59, 660–665.
- 529 Bartoń, K. (2009). *MuMIn : multi-model inference*.
- Bates, D., Mächler, M., Bolker, B. & Walker, S. (2015). Fitting Linear Mixed-Effects Models Using
 Ime4. J. Stat. Softw., 67, 1–48.
- Beisner, B.E., Haydon, D.T. & Cuddington, K. (2003). Alternative stable states in ecology. *Front. Ecol. Environ.*, 1, 376–382.
- Bharath, S., Borer, E.T., Biederman, L.A., Blumenthal, D.M., Fay, P.A., Gherardi, L.A., *et al.*(2020). Nutrient addition increases grassland sensitivity to droughts. *Ecology*, 101, e02981.
- Borer, E.T., Harpole, W.S., Adler, P.B., Lind, E.M., Orrock, J.L., Seabloom, E.W., *et al.* (2014).
 Finding generality in ecology: a model for globally distributed experiments. *Methods Ecol. Evol.*, 5, 65–73.
- 540 Braakhekke, W.G. & Hooftman, D.A.P. (1999). The resource balance hypothesis of plant species 541 diversity in grassland. *J. Veg. Sci.*, 10, 187–200.
- Brown, B.L., Downing, A.L. & Leibold, M.A. (2016). Compensatory dynamics stabilize aggregate
 community properties in response to multiple types of perturbations. *Ecology*, 97,
 2021–2033.
- 545 Cardinale, B.J., Duffy, J.E., Gonzalez, A., Hooper, D.U., Perrings, C., Venail, P., *et al.* (2012).
 546 Biodiversity loss and its impact on humanity. *Nature*, 486, 59–67.
- 547 Cardinale, B.J., Gross, K., Fritschie, K., Flombaum, P., Fox, J.W., Rixen, C., *et al.* (2013).
 548 Biodiversity simultaneously enhances the production and stability of community
 549 biomass, but the effects are independent. *Ecology*, 94, 1697–1707.
- Carnus, T., Finn, J.A., Kirwan, L. & Connolly, J. (2014). Assessing the relationship between
 biodiversity and stability of ecosystem function is the coefficient of variation always
 the best metric? *Ideas Ecol. Evol.*, 7.
- Carpenter, S.R. & Brock, W.A. (2006). Rising variance: a leading indicator of ecological
 transition. *Ecol. Lett.*, 9, 311–318.
- Carroll, I.T., Cardinale, B.J. & Nisbet, R.M. (2011). Niche and fitness differences relate the
 maintenance of diversity to ecosystem function. *Ecology*, 92, 1157–1165.
- 557 Chesson, P. (2000). Mechanisms of Maintenance of Species Diversity. *Annu. Rev. Ecol. Syst.*, 31,
 558 343–366.
- Danger, M., Daufresne, T., Lucas, F., Pissard, S. & Lacroix, G. (2008). Does Liebig's Law of the
 Minimum Scale up from Species to Communities? *Oikos*, 117, 1741–1751.
- Donohue, I., Hillebrand, H., Montoya, J.M., Petchey, O.L., Pimm, S.L., Fowler, M.S., *et al.* (2016).
 Navigating the complexity of ecological stability. *Ecol. Lett.*, 19, 1172–1185.
- Donohue, I., Petchey, O.L., Montoya, J.M., Jackson, A.L., McNally, L., Viana, M., *et al.* (2013). On
 the dimensionality of ecological stability. *Ecol. Lett.*, 16, 421–429.

565 Droop, M.R. (1974). The nutrient status of algal cells in continuous culture. J. Mar. Biol. Assoc. 566 U. K., 54, 825–855. 567 Elser, J.J., Bracken, M.E.S., Cleland, E.E., Gruner, D.S., Harpole, W.S., Hillebrand, H., et al. (2007). 568 Global analysis of nitrogen and phosphorus limitation of primary producers in 569 freshwater, marine and terrestrial ecosystems. Ecol. Lett., 10, 1135–1142. 570 Fay, P.A., Prober, S.M., Harpole, W.S., Knops, J.M.H., Bakker, J.D., Borer, E.T., et al. (2015). 571 Grassland productivity limited by multiple nutrients. Nat. Plants, 1, 15080. 572 Gilbert, B., MacDougall, A.S., Kadoya, T., Akasaka, M., Bennett, J.R., Lind, E.M., et al. (2020). 573 Climate and local environment structure asynchrony and the stability of primary 574 production in grasslands. Glob. Ecol. Biogeogr., 29, 1177–1188. 575 Gounand, I., Little, C.J., Harvey, E. & Altermatt, F. (2020). Global quantitative synthesis of 576 ecosystem functioning across climatic zones and ecosystem types. Glob. Ecol. Biogeogr., 577 29, 1139–1176. 578 Harpole, W.S., Ngai, J.T., Cleland, E.E., Seabloom, E.W., Borer, E.T., Bracken, M.E.S., et al. 579 (2011). Nutrient co-limitation of primary producer communities. Ecol. Lett., 14, 852– 580 862. 581 Harpole, W.S., Sullivan, L.L., Lind, E.M., Firn, J., Adler, P.B., Borer, E.T., et al. (2016). Addition of 582 multiple limiting resources reduces grassland diversity. *Nature*, 537, 93–96. 583 Harpole, W.S., Sullivan, L.L., Lind, E.M., Firn, J., Adler, P.B., Borer, E.T., et al. (2017). Out of the 584 shadows: multiple nutrient limitations drive relationships among biomass, light and 585 plant diversity. Funct. Ecol., 31, 1839–1846. 586 Harpole, W.S. & Tilman, D. (2006). Non-neutral patterns of species abundance in grassland 587 communities. Ecol. Lett., 9, 15–23. 588 Harpole, W.S. & Tilman, D. (2007). Grassland species loss resulting from reduced niche 589 dimension. Nature, 446, 791-793. 590 Hautier, Y., Niklaus, P.A. & Hector, A. (2009). Competition for Light Causes Plant Biodiversity 591 Loss After Eutrophication. Science, 324, 636–638. 592 Hautier, Y., Seabloom, E.W., Borer, E.T., Adler, P.B., Harpole, W.S., Hillebrand, H., et al. (2014). 593 Eutrophication weakens stabilizing effects of diversity in natural grasslands. Nature, 508, 594 521-525. 595 Hautier, Y., Tilman, D., Isbell, F., Seabloom, E.W., Borer, E.T. & Reich, P.B. (2015). Anthropogenic 596 environmental changes affect ecosystem stability via biodiversity. Science, 348, 336-597 340. 598 Hautier, Y., Zhang, P., Loreau, M., Wilcox, K.R., Seabloom, E.W., Borer, E.T., et al. (2020). 599 General destabilizing effects of eutrophication on grassland productivity at multiple 600 spatial scales. Nat. Commun., 11, 5375. 601 Hector, A., Hautier, Y., Saner, P., Wacker, L., Bagchi, R., Joshi, J., et al. (2010). General stabilizing 602 effects of plant diversity on grassland productivity through population asynchrony and 603 overyielding. *Ecology*, 91, 2213–2220. 604 Hedges, L.V., Gurevitch, J. & Curtis, P.S. (1999). The Meta-Analysis of Response Ratios in 605 Experimental Ecology. Ecology, 80, 1150–1156. Hijmans, R.J., Cameron, S.E., Parra, J.L., Jones, P.G. & Jarvis, A. (2005). Very high resolution 606 607 interpolated climate surfaces for global land areas. Int. J. Climatol., 25, 1965–1978.

- Hou, E., Luo, Y., Kuang, Y., Chen, C., Lu, X., Jiang, L., *et al.* (2020). Global meta-analysis shows
 pervasive phosphorus limitation of aboveground plant production in natural terrestrial
 ecosystems. *Nat. Commun.*, 11, 637.
- Huxman, T.E., Smith, M.D., Fay, P.A., Knapp, A.K., Shaw, M.R., Loik, M.E., *et al.* (2004).
 Convergence across biomes to a common rain-use efficiency. *Nature*, 429, 651–654.
- Isbell, F., Reich, P.B., Tilman, D., Hobbie, S.E., Polasky, S. & Binder, S. (2013). Nutrient
 enrichment, biodiversity loss, and consequent declines in ecosystem productivity. *Proc. Natl. Acad. Sci.*, 110, 11911–11916.
- 616 Ives, A.R. & Carpenter, S.R. (2007). Stability and Diversity of Ecosystems. *Science*, 317, 58–62.
- 617 Ives, A.R., Klug, J.L. & Gross, K. (2000). Stability and species richness in complex communities.
 618 *Ecol. Lett.*, 3, 399–411.
- Kaspari, M. & Powers, J.S. (2016). Biogeochemistry and Geographical Ecology: Embracing All
 Twenty-Five Elements Required to Build Organisms. *Am. Nat.*, 188, S62–S73.
- Kaspari, M., Roeder, K.A., Benson, B., Weiser, M.D. & Sanders, N.J. (2017). Sodium co-limits and
 catalyzes macronutrients in a prairie food web. *Ecology*, 98, 315–320.
- Koerner, S.E., Avolio, M.L., Pierre, K.J.L., Wilcox, K.R., Smith, M.D. & Collins, S.L. (2016). Nutrient
 additions cause divergence of tallgrass prairie plant communities resulting in loss of
 ecosystem stability. *J. Ecol.*, 104, 1478–1487.
- Kohli, M., Borer, E.T., Kinkel, L. & Seabloom, E.W. (2019). Stability of grassland production is
 robust to changes in the consumer food web. *Ecol. Lett.*, 22, 707–716.
- Kremen, C. (2005). Managing ecosystem services: what do we need to know about their
 ecology? *Ecol. Lett.*, 8, 468–479.
- LeBauer, D.S. & Treseder, K.K. (2008). Nitrogen Limitation of Net Primary Productivity in
 Terrestrial Ecosystems Is Globally Distributed. *Ecology*, 89, 371–379.
- Lepš, J., Götzenberger, L., Valencia, E. & Bello, F. de. (2019). Accounting for long-term
 directional trends on year-to-year synchrony in species fluctuations. *Ecography*, 42,
 1728–1741.
- Levin, S.A. (1970). Community Equilibria and Stability, and an Extension of the Competitive
 Exclusion Principle. *Am. Nat.*, 104, 413–423.
- Loreau, M. (2010). From Populations to Ecosystems: Theoretical Foundations for a New
 Ecological Synthesis (MPB-46). Princeton University Press.
- Loreau, M. & de Mazancourt, C. (2008). Species Synchrony and Its Drivers: Neutral and
 Nonneutral Community Dynamics in Fluctuating Environments. *Am. Nat.*, 172, E48–E66.
- MacDougall, A.S., McCann, K.S., Gellner, G. & Turkington, R. (2013). Diversity loss with
 persistent human disturbance increases vulnerability to ecosystem collapse. *Nature*,
 494, 86–89.
- McCann, K.S., Cazelles, K., MacDougall, A.S., Fussmann, G.F., Bieg, C., Cristescu, M., *et al.*(2021). Landscape modification and nutrient-driven instability at a distance. *Ecol. Lett.*,
 24, 398–414.
- Mellinger, M.V. & McNaughton, S.J. (1975). Structure and Function of Successional Vascular
 Plant Communities in Central New York. *Ecol. Monogr.*, 45, 161–182.
- Morgan, J.W., Dwyer, J.M., Price, J.N., Prober, S.M., Power, S.A., Firn, J., *et al.* (2016). Species
 origin affects the rate of response to inter-annual growing season precipitation and
 nutrient addition in four Australian native grasslands. *J. Veg. Sci.*, 27, 1164–1176.

- Muraina, T.O., Xu, C., Yu, Q., Yang, Y., Jing, M., Jia, X., *et al.* (2021). Species asynchrony stabilises
 productivity under extreme drought across Northern China grasslands. *J. Ecol.*, 109,
 1665–1675.
- Nakagawa, S. & Cuthill, I.C. (2007). Effect size, confidence interval and statistical significance: a
 practical guide for biologists. *Biol. Rev.*, 82, 591–605.
- Paruelo, J.M., Lauenroth, W.K., Burke, I.C. & Sala, O.E. (1999). Grassland Precipitation-Use
 Efficiency Varies Across a Resource Gradient. *Ecosystems*, 2, 64–68.
- R Foundation for Statistical Computing. (n.d.). *R: The R Project for Statistical Computing*. Vienna,
 Austria.
- Rosenzweig, M.L. (1971). Paradox of Enrichment: Destabilization of Exploitation Ecosystems in
 Ecological Time. *Science*, 171, 385–387.
- Scheffer, M., Carpenter, S., Foley, J.A., Folke, C. & Walker, B. (2001). Catastrophic shifts in
 ecosystems. *Nature*, 413, 591–596.
- Seabloom, E.W., Adler, P.B., Alberti, J., Biederman, L., Buckley, Y.M., Cadotte, M.W., *et al.*(2021). Increasing effects of chronic nutrient enrichment on plant diversity loss and
 ecosystem productivity over time. *Ecology*, 102, e03218.
- Tilman, D. (1982). *Resource competition and community structure*. Monographs in population
 biology. Princeton University Press, Princeton, N.J.
- Tilman, D. (1996). Biodiversity: Population Versus Ecosystem Stability. *Ecology*, 77, 350–363.
- Tilman, D. (1999). The Ecological Consequences of Changes in Biodiversity: A Search for General
 Principles. *Ecology*, 80, 1455–1474.
- Tilman, D. (2001). Diversity and Productivity in a Long-Term Grassland Experiment. *Science*, 294,
 843–845.
- Tilman, D. & Lehman, C. (2001). Human-caused environmental change: Impacts on plant
 diversity and evolution. *Proc. Natl. Acad. Sci.*, 98, 5433–5440.
- Tilman, D., Reich, P.B. & Knops, J.M.H. (2006). Biodiversity and ecosystem stability in a decade long grassland experiment. *Nature*, 441, 629–632.
- Valencia, E., de Bello, F., Galland, T., Adler, P.B., Lepš, J., E-Vojtkó, A., *et al.* (2020a). Synchrony
 matters more than species richness in plant community stability at a global scale. *Proc. Natl. Acad. Sci.*, 117, 24345–24351.
- Valencia, E., Bello, F. de, Lepš, J., Galland, T., E-Vojtkó, A., Conti, L., *et al.* (2020b). Directional
 trends in species composition over time can lead to a widespread overemphasis of yearto-year asynchrony. *J. Veg. Sci.*, 31, 792–802.
- Wang, J., Knops, J.M.H., Brassil, C.E. & Mu, C. (2017). Increased productivity in wet years drives
 a decline in ecosystem stability with nitrogen additions in arid grasslands. *Ecology*, 98,
 1779–1786.
- 688

689 FIGURES



690 691 Figure 1. Effect of seven years of single- and multiple-nutrient additions on the detrended 692 stability (a), temporal mean (b), and detrended standard deviation (SD) (c) of grassland biomass 693 production in 34 sites. Different individual and combined additions of nitrogen (N), phosphorus 694 (P), and potassium with essential nutrients (K) influenced the magnitude of stability, mean and 695 variability responses. The points show percent change relative to unenriched control plots. They 696 represent back-transformed fixed effect estimates (with 95% confidence intervals) from mixed 697 effects models that accounted for site as a random grouping factor. See Supplementary 698 Information for detailed model specification and summary statistics.



Figure 2. Relationship between change in mean biomass production and change in stability under
 seven years of different individual and combined additions of nitrogen (N), phosphorus (P), and
 potassium with essential nutrients (K). Larger biomass responses were associated with larger
 declines in stability for all treatments except NP and NPK. Points show plots within blocks at 34
 grassland sites. Colored lines are fixed-effect regression slopes for each treatment from mixed
 effects models. See Supplementary Information for detailed model specification and summary
 statistics.



Figure 3. Effect of individual and combined additions of nitrogen (N), phosphorus (P), and potassium with essential nutrients (K) on the temporal mean of species richness (a) and detrended species synchrony (b) in 34 grassland sites. Points show fixed effect estimates with 95% confidence intervals (see Fig. 1 caption and Supplementary Information).



Figure 4. Nutrient-driven changes in species richness contribute to nutrient effects on the temporal stability (a), mean (b), and standard deviation (SD) (c) of grassland biomass production. Points show plots within blocks at 34 globally distributed grassland sites. Colored lines are fixedeffect regression slopes from mixed effects models for individual and combined nitrogen (N), phosphorus (P), and potassium with essential nutrients (K) addition treatments. Similar trends caused overlapping lines for some treatments in panels a (PK overlaps P) and c (PK overlaps K). See Supplementary Information for detailed model specification and summary statistics.





Figure 5. Nutrient-driven changes in species synchrony contribute to nutrient effects on the
stability of grassland biomass production over seven years. Colored lines show fixed-effect slopes
within different nitrogen (N), phosphorus (P), and potassium with essential nutrients (K)
treatments (see Fig. 4 caption and Supplementary Information).

735 SUPPLEMENTARY INFORMATION

Table S1. Summary of site characteristics for the 34 sites included in the study. Sites were located
in 9 countries across 6 continents. Sites typically had 3 experimental blocks, but this ranged from
1-6. The sites span a broad gradient of site-level plant species richness, mean annual precipitation
(MAP; mm), and mean annual temperature (MAT; °C). Data beyond seven years were removed
where applicable. For sites with missing data, the available years of treatment data that were

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Site	Country	Continent	Lat.	Long.	Elevation	Blocks	Richness	MAP	MAT	Years
bldr.us	US	N. America	40.0	-105.2	1633	2	49	487	9.9	1-7
bnch.us	US	N. America	44.3	-122.0	1318	3	61	1618	6.8	1-7
bogong.au	AU	Australia	-36.9	147.3	1760	3	51	1678	6	1-7
burrawan.au	AU	Australia	-27.7	151.1	425	3	45	643	18	1-7
cbgb.us	US	N. America	41.8	-93.4	275	6	113	871	9.3	1-7
cdcr.us	US	N. America	45.4	-93.2	270	5	142	740	6.3	1-7
cdpt.us	US	N. America	41.2	-101.6	965	6	85	456	9.6	1-7
comp.pt	PT	Europe	38.8	-8.8	200	3	105	564	17	1-7
cowi.ca	CA	N. America	48.8	-123.6	50	3	29	762	10	1-7
elliot.us	US	N. America	32.9	-117.1	200	3	51	344	18	1-7
frue.ch	CH	Europe	47.1	8.5	995	3	40	1546	7	1-7
hall.us	US	N. America	36.9	-86.7	194	3	53	1289	14	1-7
hopl.us	US	N. America	39.0	-123.1	598	3	132	1065	13	1-7
kibber.in	IN	Asia	32.3	78.0	4241	3	37	400	-1.5	1-4,6
kiny.au	AU	Australia	-36.2	143.8	90	3	114	408	16	1-7
koffler.ca	CA	N. America	44.0	-79.5	301	3	51	853	6.3	1-7
konz.us	US	N. America	39.1	-96.6	440	3	109	889	12	1-5,7
lancaster.uk	UK	Europe	54.0	-2.6	180	3	34	1522	8	1-3,
										6,7
look.us	US	N. America	44.2	-122.1	1500	3	62	1877	6.9	1-7
marc.ar	AR	S. America	-37.7	-57.4	6	3	67	907	14	1-7
mcla.us	US	N. America	38.9	-122.4	642	3	88	936	14	1-7
mtca.au	AU	Australia	-31.8	117.6	285	4	71	324	18	1-7
saline.us	US	N. America	39.0	-99.1	440	3	122	608	12	1-7
sedg.us	US	N. America	34.7	-120.0	550	3	34	478	16	1-7
sevi.us	US	N. America	34.4	-106.7	1600	1	85	252	13	1-7
sgs.us	US	N. America	40.8	-104.8	1650	3	78	369	8.9	1-7
shps.us	US	N. America	44.2	-112.2	910	4	97	246	5.3	1-5
sier.us	US	N. America	39.2	-121.3	197	5	128	936	16	1-7
smith.us	US	N. America	48.2	-122.6	62	3	66	605	10	1-5
spin.us	US	N. America	38.1	-84.5	271	3	52	1152	13	1-7
temple.us	US	N. America	31.0	-97.3	184	3	99	877	19	1-7
trel.us	US	N. America	40.1	-88.8	200	3	25	992	11	1-7
ukul.za	ZA	Africa	-29.7	30.4	842	3	176	832	18	1-7
valm.ch	CH	Europe	46.6	10.4	2320	3	122	681	0.1	1-7

Table S2. Assessment of the sensitivity of detrended stability effect size (LRRs) estimates from model 1 to the number of years assessed. We analyzed data for seven treatment years. This choice impacted the length of time available for temporal dynamics to unfold as well as the number of sites it was possible to use in the analysis. Significant effect sizes at alpha 0.05 are shown in bold and underlined. Effect sizes with P values below alpha 0.1 are underlined. The effects presented in the paper were most representative of those seen in subsets of between 1-6 and 1-9 years of treatment, which included from 43 to 28 sites, respectively.

			Nutrient Treatment						
Years	Sites	Parameter	К	Р	N	РК	NK	NP	NPK
3	70	Estimate	0.028	0.059	0.001	0.096	0.076	0.077	-0.034
		Р	0.642	0.335	0.986	0.115	0.212	0.208	0.573
4	58	Estimate	-0.010	-0.055	<u>-0.119</u>	0.019	-0.045	0.011	-0.078
		Р	0.853	0.312	<u>0.030</u>	0.732	0.415	0.837	0.155
5	51	Estimate	-0.024	-0.059	<u>-0.083</u>	-0.017	-0.058	-0.003	-0.076
		Р	0.625	0.238	<u>0.099</u>	0.741	0.245	0.956	0.128
6	43	Estimate	-0.028	<u>-0.078</u>	<u>-0.104</u>	-0.046	<u>-0.122</u>	-0.035	<u>-0.087</u>
		Р	0.541	<u>0.093</u>	<u>0.025</u>	0.321	<u>0.009</u>	0.455	<u>0.061</u>
7	34	Estimate	-0.044	<u>-0.115</u>	<u>-0.163</u>	-0.049	<u>-0.165</u>	<u>-0.135</u>	<u>-0.129</u>
		Р	0.322	<u>0.011</u>	<0.001	0.278	<u><0.001</u>	<u>0.003</u>	<u>0.004</u>
8	29	Estimate	-0.026	<u>-0.092</u>	<u>-0.135</u>	-0.061	<u>-0.140</u>	<u>-0.109</u>	<u>-0.122</u>
		Р	0.573	<u>0.051</u>	<u>0.004</u>	0.194	<u>0.003</u>	<u>0.021</u>	<u>0.010</u>
9	28	Estimate	-0.032	<u>-0.071</u>	<u>-0.122</u>	-0.048	<u>-0.114</u>	<u>-0.107</u>	<u>-0.119</u>
		Р	0.444	<u>0.090</u>	<u>0.004</u>	0.255	<u>0.007</u>	<u>0.011</u>	<u>0.005</u>
10	21	Estimate	-0.001	-0.034	-0.035	0.006	<u>-0.090</u>	-0.030	-0.017
		Р	0.979	0.484	0.472	0.908	0.064	0.531	0.722
11	16	Estimate	0.030	-0.037	-0.050	0.004	<u>-0.113</u>	-0.053	-0.069
		Р	0.547	0.463	0.325	0.934	<u>0.027</u>	0.296	0.175
12	6	Estimate	-0.025	-0.007	-0.070	-0.036	-0.087	-0.052	-0.114
		Ρ	0.685	0.915	0.258	0.561	0.158	0.399	0.066

752	Table S3. Effect sizes of the temporal mean, standard deviation and stability of grassland biomass
753	production in response to well-established mechanisms of function and stability. We present this
754	range of log response ratios (LRR) to aid interpretation of the biological significance of our results.
755	Here, the effect sizes denote the impact of the listed experimental manipulation relative to
756	control conditions within the study. The references used to provide these benchmarks include
757	meta-analysis of multiple studies of stability effects (1), as well as assessment of sites with a wide
758	geographical distribution (2).

Response	LRR	% change	Examples from previous grassland studies	Ref*
Biomass	0.22	25	LRR = 0.24: N enrichment - 34 kg ha ⁻¹ / 3.4 g m ² LRR = 0.27: Benchmark for biologically significant	(1)
			mean biomass response	(2)
	0.4	50	LRR = 0.57: N enrichment - 54 kg ha ⁻¹ / 5.4 g m ²	(3)
	0.56	75	LRR = 0.66: N enrichment - 270 kg ha ⁻¹ / 27 g m ²	(1)
			LRR = 0.74: Increasing species richness from 1 to 16	(2)
SD	0.22	25	LRR = 0.24: Reducing species richness from 16 to 4	(1)
	0.4	50	LRR = 0.48: N enrichment - 54 kg ha ⁻¹ / 5.4 g m ² LRR = 0.48: Reducing species richness from 16 to 2	(1)
	0.56	75	LRR = 0.60: Reducing species richness from 16 to 1	(1)
S _d	-0.11	-10	LRR = -0.11: N enrichment - 54 kg ha ⁻¹ / 5.4 g m ²	(1)
	-0.22	-20	LRR = 0.22: Reducing species richness from 16 to 2	(1)
			LRR = 0.29: N enrichment - 270 kg ha ⁻¹ / 27 g m ²	(1)
	-0.43	-35	LRR = 0.41: Reducing species richness from 16 to 1	(1)

* (1): Hautier et al., 2015; (2): Fay et al., 2015; (3): Tilman et al., 2012

```
761

        Table S4. Model summary for nutrient treatment effects on stability

762
763
      Linear mixed model fit by REML. t-tests use Satterthwaite's method
764
     ['lmerModLmerTest']
765
      Formula: lrr.detr.s ~ 0 + trt + (1 | site code)
766
         Data: lrr
767
768
      REML criterion at convergence: 886.6
769
770
      Scaled residuals:
771
          Min
                   1Q Median
                               30
                                           Max
772
      -3.7009 -0.6123 0.0431 0.5866 3.2800
773
774
     Random effects:
775
     Groups Name
                             Variance Std.Dev.
776
      site code (Intercept) 0.01586 0.1259
777
     Residual
                            0.16906 0.4112
778
      Number of obs: 777, groups: site code, 34
779
780
      Fixed effects:
781
             Estimate Std. Error
                                          df t value Pr(>|t|)
              -0.04438 0.04474 232.60035 -0.992 0.322231
782
     trtK
783
             -0.16315 0.04474 232.60035 -3.647 0.000328 ***
     trtN
784
     trtNK -0.16495 0.04474 232.60035 -3.687 0.000283 ***
     trtNP-0.135400.04474232.60035-3.0260.002753**trtNPK-0.129040.04474232.60035-2.8840.004291**trtP-0.115040.04474232.60035-2.5710.010752*
785
786
787
788
     trtPK -0.04869 0.04474 232.60035 -1.088 0.277538
789
      ____
790
      Signif. codes: 0 ,Äò***,Äô 0.001 ,Äò**,Äô 0.01 ,Äò*,Äô 0.05 ,Äò.,Äô 0.1
791
      ,Äò ,Äô 1
792
793
      Correlation of Fixed Effects:
794
             trtK trtN trtNK trtNP trtNPK trtP
795
             0.239
     trtN
796
     trtNK 0.239 0.239
797
     trtNP 0.239 0.239 0.239
798
     trtNPK 0.239 0.239 0.239 0.239
799
     trtP 0.239 0.239 0.239 0.239 0.239
800
     trtPK 0.239 0.239 0.239 0.239 0.239 0.239
801
```

803 804 Stability of biomass: 805 806 lrr.detr.s ~ 0 + N * P * K + (1 | site_code) 807 808 Nutrient Estimate SE df t.value Ρ 809 N -0.163 0.045 232.600 -3.647 0.000 -0.115 0.045 232.600 -2.571 0.011 810 Ρ К -0.044 0.045 232.600 -0.992 0.322 811 812 N:P 0.143 0.071 628.152 2.010 0.045 813 N:K 0.043 0.071 628.152 0.599 0.549 0.111 0.071 628.152 1.558 0.120 814 P:K 815 N:P:K -0.103 0.106 762.413 -0.972 0.331 816 817 818 Mean biomass: 819 820 lrr.mean \sim 0 + N * P * K + (1 | site code) 821 822 Nutrient Estimate SE df t.value Ρ 823 N 0.194 0.039 111.782 5.003 0.000 824 0.056 0.039 111.782 1.453 0.149 Ρ 825 к -0.004 0.039 111.782 -0.094 0.926 N:P 0.135 0.056 361.270 2.421 0.016 826 827 N:K 0.065 0.056 361.270 1.161 0.246 828 P:K 0.100 0.056 361.270 1.796 0.073 829 N:P:K -0.139 0.080 655.989 -1.740 0.082 830 831 832 SD of biomass: 833 834 lrr.detr.sd \sim 0 + N * P * K + (1 | site code) 835 836 Nutrient Estimate SE df t.value Ρ 837 N 0.359 0.063 198.778 5.688 0.000 838 0.173 0.063 198.778 2.745 0.007 Ρ 839 0.043 0.063 198.778 0.675 0.500 K 840 N:P -0.009 0.098 577.917 -0.096 0.924 N:K 0.020 0.098 577.917 0.207 0.836 P:K -0.012 0.098 577.917 -0.125 0.901 841 842 843 N:P:K -0.034 0.145 752.494 -0.235 0.814 844

```
846
847
      Linear mixed model fit by REML. t-tests use Satterthwaite's method [
848
      lmerModLmerTest]
849
      Formula: lrr.mean ~ 0 + trt + (1 | site code)
850
         Data: lrr
851
852
      REML criterion at convergence: 423.3
853
854
      Scaled residuals:
855
          Min 1Q Median 3Q
                                            Max
856
      -3.1935 -0.6635 0.0369 0.5792 5.4567
857
858
     Random effects:
      Groups Name Variance Std.Dev.
859
860
      site code (Intercept) 0.02348 0.1532
861
      Residual 0.08955 0.2992
862
      Number of obs: 777, groups: site code, 34
863
864
      Fixed effects:
                                              df t value
              Estimate Std. Error
865
                                                                        Pr(>|t|)

        -0.003031
        0.038804
        111.782393
        -0.094
        0.92561

        0.194117
        0.038804
        111.782393
        5.003
        0.00000211930
        ***

866
     trtK
867
     trtN
     trtNK0.2553240.038804111.7823936.5800.00000000158***trtNP0.3857400.038804111.7823939.941< 0.000000000000002</td>***
868
869
870
     trtNPK 0.408658 0.038804 111.782393 10.531 < 0.00000000000002 ***
     trtP0.0563910.038804111.7823931.453trtPK0.1530580.038804111.7823933.944
871
                                                                         0.14896
872
                                                                          0.00014 ***
873
874
      Signif. codes: 0 ,Äò***,Äô 0.001 ,Äò**,Äô 0.01 ,Äò*,Äô 0.05 ,Äò.,Äô 0.1
875
      ,Äò ,Äô 1
876
877
      Correlation of Fixed Effects:
878
            trtK trtN trtNK trtNP trtNPK trtP
879
      trtN 0.464
880
      trtNK 0.464 0.464
     trtNP 0.464 0.464 0.464
881
882
     trtNPK 0.464 0.464 0.464 0.464
883
      trtP 0.464 0.464 0.464 0.464 0.464
884
      trtPK 0.464 0.464 0.464 0.464 0.464 0.464
885
886
```

Table S6. Model summary for nutrient treatment effects on mean biomass production

887 **Table S7.** Model summary for nutrient treatment effects on biomass temporal variability

```
888
889
      Linear mixed model fit by REML. t-tests use Satterthwaite's method [
890
      lmerModLmerTest]
891
      Formula: lrr.detr.sd ~ 0 + trt + (1 | site code)
892
         Data: lrr
893
894
      REML criterion at convergence: 1372.5
895
896
      Scaled residuals:
897
          Min 10 Median 30
                                             Max
898
      -3.2911 -0.6016 -0.0416 0.6161 4.2848
899
900
     Random effects:
901
      Groups Name Variance Std.Dev.
902
      site code (Intercept) 0.03809 0.1952
903
      Residual 0.31538 0.5616
904
      Number of obs: 777, groups: site code, 34
905
906
      Fixed effects:
              Estimate Std. Error df t value
0.04262 0.06313 198.77816 0.675
907
                                                                      Pr(>|t|)
908
                                                                      0.50042
     trtK
909
              0.35914 0.06313 198.77816 5.688 0.00000004537297638 ***
      trtN
     trtNK0.422140.06313198.778166.6860.0000000022616030***trtNP0.523010.06313198.778168.2840.000000000000001739***trtNPK0.539570.06313198.778168.5460.0000000000000334***trtP0.173300.06313198.778162.7450.00661**trtPK0.203620.06313198.778163.2250.00147**
910
911
912
913
914
915
916
      Signif. codes: 0 ,Äò***,Äô 0.001 ,Äò**,Äô 0.01 ,Äò*,Äô 0.05 ,Äò.,Äô 0.1
917
      ,Äò ,Äô 1
918
919
      Correlation of Fixed Effects:
920
            trtK trtN trtNK trtNP trtNPK trtP
921
     trtN 0.287
922
      trtNK 0.287 0.287
923
     trtNP 0.287 0.287 0.287
     trtNPK 0.287 0.287 0.287 0.287
924
925
     trtP 0.287 0.287 0.287 0.287 0.287
926
      trtPK 0.287 0.287 0.287 0.287 0.287 0.287
927
```

928	Table S8. Number of grassland sites (total = 34) at which predicted biomass and stability
929	responses to nutrient inputs exceeded effects demonstrated elsewhere in the literature. See the
930	text and Table S2 for a description of comparable effect sizes.

_		Me	Mean biomass		Standa	Standard dev. (σ_{resid})			CV_{detr}^{-1}	
	LRR:	0.22	0.4	0.56	0.22	0.4	0.56	-0.11	-0.22	-0.43
	% Change:	25	50	75	25	50	75	-10	-20	-35
	К	1	0	0	5	0	0	8	0	0
	Р	4	0	0	12	3	0	19	7	0
	Ν	13	1	0	25	12	3	23	8	0
	РК	9	1	0	13	3	0	8	0	0
	NK	21	4	1	28	16	7	24	8	0
	NP	30	15	4	33	25	13	21	8	0
_	NPK	31	18	5	33	26	15	20	8	0

932 **Table S9.** Model summary for relationship between mean biomass response and stability

933 response:

934

935 Linear mixed model fit by maximum likelihood . t-tests use Satterthwaite's 936 method ['lmerModLmerTest'] 937 Formula: lrr.detr.s ~ trt:lrr.mean + (1 | site code) 938 Data: lrr 939 940 AIC BIC logLik deviance df.resid 941 876.5 -405.0 829.9 809.9 767 942 943 Scaled residuals: 944 Min 1Q Median 3Q Max 945 -3.5586 -0.6277 0.0145 0.6153 3.5164 946 947 Random effects: 948 Groups Name Variance Std.Dev. 949 site code (Intercept) 0.01723 0.1313 950 Residual 0.15732 0.3966 951 Number of obs: 777, groups: site_code, 34 952 953 Fixed effects: 954 Estimate Std. Error df t value Pr(>|t|) 955 -0.06156 0.02849 41.51052 -2.161 (Intercept) 0.03652 * 956 trtK:lrr.mean -0.31256 0.11053 765.49997 -2.828 0.00481 ** trtN:lrr.mean-0.311290.10190767.26283-3.0550.00233**trtNK:lrr.mean-0.365720.08749765.25658-4.1800.000032498*** 957 958 959 trtNP:lrr.mean -0.13353 0.08195 765.31866 -1.629 0.10363 trtNPK:lrr.mean-0.145340.07404766.20068-1.963trtP:lrr.mean-0.259050.12032768.48178-2.153 960 0.05002 . 961 0.03163 * trtPK:lrr.mean -0.54368 0.10256 767.07393 -5.301 0.000000151 *** 962 963 ___ 964 Signif. codes: 0 ,Äò***,Äô 0.001 ,Äò**,Äô 0.01 ,Äò*,Äô 0.05 ,Äò.,Äô 0.1 965 ,Äò ,Äô 1 966 967 Correlation of Fixed Effects: 968 (Intr) trtK:. trtN:. trNK:. trNP:. tNPK:. trtP:. 969 trtK:lrr.mn -0.020 970 trtN:lrr.mn -0.152 0.030 971 trtNK:lrr.m -0.165 0.031 0.101 972 trtNP:lrr.m -0.214 0.024 0.114 0.117 trtNPK:lrr. -0.206 0.023 0.115 0.117 0.146 973 974 trtP:lrr.mn -0.073 0.032 0.057 0.057 0.061 0.064 975 trtPK:lrr.m -0.132 0.022 0.071 0.074 0.094 0.093 0.045 976 977

Table S10. Model averaging summaries for drivers of stability, biomass and biomass variability

```
979
      responses.
980
981
      Stability of biomass:
982
983
      Maximum model:
984
      lrr.detr.s ~ trt + trt * (lrr.rich + lrr.sync + log.map + log.map.var) +
985
          (1 | site code)
986
987
      Retained fixed effects after model selection (delta <4):
988
                          lrr.rich lrr.sync log.map log.map.var trt
989
                                   0.93
                                            0.28
                                                 0.24
                                                               0.07
      Sum of weights:
                          1.00
990
      N containing models:
                             6
                                      5
                                               2
                                                       2
                                                                  1
991
992
      Full average:
993
                  Estimate SE
                                Adj.SE z
                                             Ρ
994
      (Intercept) -0.012 0.253 0.253 0.048 0.961
995
                          0.066 0.066 2.624 0.009
                  0.173
      lrr.rich
                         0.017 0.017 1.944 0.052
996
                  -0.032
      lrr.sync
997
      log.map
                 -0.011 0.033 0.033 0.327 0.744
998
      log.map.var -0.004 0.027 0.027 0.142 0.887
999
                  -0.008
                         0.031 0.031 0.249 0.803
      trtN
1000
      trtNK
                  -0.009 0.034 0.034 0.255 0.799
1001
                -0.006 0.026 0.026 0.231 0.818
      trtNP
1002
                 -0.005
                          0.023 0.023 0.215 0.830
      trtNPK
1003
      trtP
                  -0.006
                          0.025 0.025 0.229 0.819
1004
      trtPK
                  0.000
                         0.015 0.015 0.008 0.994
1005
1006
      Conditional average:
1007
                  Estimate SE
                                Adj.SE z
                                             Ρ
1008
      (Intercept) -0.012
                         0.253 0.253 0.048 0.961
1009
                          0.066 0.066 2.624 0.009
      lrr.rich
                  0.173
1010
      lrr.sync
                  -0.035 0.015 0.015 2.383 0.017
1011
      log.map
                  -0.039 0.054 0.054 0.735 0.463
1012
      log.map.var -0.016
                         0.054 0.054 0.301 0.763
1013
                 -0.105
                         0.055 0.055 1.930 0.054
      trtN
1014
                  -0.118
                         0.055 0.055 2.163 0.031
      trtNK
1015
                  -0.080
      trtNP
                          0.055 0.055 1.467 0.142
                 -0.067
1016
      trtNPK
                         0.055 0.055 1.216 0.224
1017
      trtP
                  -0.078 0.054 0.054 1.429 0.153
1018
                  -0.002 0.054 0.054 0.028 0.977
      trtPK
1019
1020
```

Ma 'n n nalal							
Maximum model:	/ [*]		. 1				
(1 site_co	ode)	LTT•TT(JII T II.	r.sync	- 10g.m	ap - 109.1	llap•va
Retained fixed e	effects a:	fter mo	odel sei	lection	n (delta	<4):	
	lrr	.rich t	trt lr:	r.sync	log.map	log.map.v	Jar
lrr.rich:trt				-	5 1	5 1	
Sum of weights:	1.00) :	1.00 0.4	46	0.35	0.23	Ο.
N containing mod	dels: 11	1	11	5	5	4	
Full average:							
	Estimate	SE	Adj.SE	Z	P		
(Intercept)	-0.117	0.282	0.282	0.414	0.679		
lrr.rich	-0.230	0.069	0.069	3.347	0.001		
trtN	0.186	0.040	0.040	4.659	0.000		
trtNK	0.246	0.040	0.040	6.112	0.000		
trtNP	0.369	0.040	0.040	9.217	0.000		
trtNPK	0.374	0.043	0.043	8.739	0.000		
trtP	0.065	0.040	0.040	1.654	0.098		
trtPK	0.148	0.040	0.040	3.717	0.000		
lrr.sync	-0.007	0.010	0.010	0.661	0.509		
log.map	0.018	0.038	0.038	0.457	0.647		
log.map.var	-0.003	0.026	0.026	0.110	0.913		
lrr.rich:trtN	0.008	0.066	0.066	0.114	0.909		
lrr.rich:trtNK	0.016	0.076	0.076	0.216	0.829		
lrr.rich:trtNP	-0.002	0.061	0.061	0.036	0.971		
lrr.rich:trtNPK	-0.038	0.115	0.115	0.330	0.742		
lrr.rich:trtP	-0.026	0.098	0.098	0.266	0.791		
lrr.rich:trtPK	-0.020	0.083	0.083	0.240	0.810		
Conditional ave:	rage:						
	Estimate	SE	Adj.SE	Z	Р		
(Intercept)	-0.117	0.282	0.282	0.414	0.679		
lrr.rich	-0.230	0.069	0.069	3.347	0.001		
trtN	0.186	0.040	0.040	4.659	0.000		
trtNK	0.246	0.040	0.040	6.112	0.000		
trtNP	0.369	0.040	0.040	9.217	0.000		
trtNPK	0.374	0.043	0.043	8.739	0.000		
trtP	0.065	0.040	0.040	1.654	0.098		
trtPK	0.148	0.040	0.040	3.717	0.000		
lrr.sync	-0.015	0.011	0.011	1.376	0.169		
log.map	0.050	0.051	0.051	0.989	0.323		
log.map.var	-0.012	0.052	0.052	0.231	0.817		
lrr.rich:trtN	0.058	0.175	0.175	0.331	0.741		
lrr.rich:trtNK	0.126	0.175	0.175	0.720	0.471		
lrr.rich:trtNP	-0.017	0.169	0.169	0.101	0.920		
lrr.rich:trtNPK	-0.291	0.166	0.166	1.748	0.080		
lrr.rich:trtP	-0.200	0.198	0.198	1.012	0.312		
lrr.rich:trtPK	-0.154	0.181	0.181	0.847	0.397		

1075 SD of biomass: 1076 1077 Maximum model: 1078 lrr.detr.sd ~ trt + trt * (lrr.rich + lrr.sync + log.map + log.map.var) + 1079 (1 | site code) 1080 1081 Retained fixed effects after model selection (delta <4): 1082 lrr.rich trt log.map lrr.sync log.map.var 1083 Sum of weights: 1.00 1.00 0.43 0.38 0.27 1084 N containing models: 8 8 4 4 4 1085 1086 Full average: 1087 Estimate SE Adj.SE z Р 1088 (Intercept) -0.235 0.485 0.486 0.484 0.628 1089 lrr.rich -0.411 0.091 0.091 4.504 0.000 1090 trtN 0.292 0.074 0.074 3.931 0.000 1091 0.358 0.074 0.074 4.809 0.000 trtNK 0.074 0.075 5.989 0.000 1092 0.446 trtNP 1093 0.443 0.075 0.075 5.898 0.000 trtNPK 1094 0.143 0.074 0.074 1.923 0.055 trtP 1095 trtPK 0.149 0.074 0.074 2.009 0.045 log.map 0.039 0.066 0.066 0.591 0.554 lrr.sync 0.008 0.016 0.016 0.506 0.613 1096 1097 1098 log.map.var 0.001 0.040 0.040 0.026 0.980 1099 1100 Conditional average: 1101 Estimate SE Adj.SE z Ρ 1102 (Intercept) -0.235 0.485 0.486 0.484 0.628 1103 lrr.rich -0.411 0.091 0.091 4.504 0.000 1104 trtN 0.292 0.074 0.074 3.931 0.000 1105 trtNK 0.358 0.074 0.074 4.809 0.000 1106 0.446 0.074 0.075 5.989 0.000 trtNP 1107 0.443 0.075 0.075 5.898 0.000 trtNPK 1108 0.143 0.074 0.074 1.923 0.055 trtP 1109 trtPK 0.149 0.074 0.074 2.009 0.045 log.map 0.091 0.074 0.074 1.232 0.218 lrr.sync 0.021 0.020 0.020 1.062 0.288 1110 log.map 1111 log.map.var 0.004 0.077 0.077 0.050 0.960 1112 1113

```
1115
1116
      Linear mixed model fit by REML. t-tests use Satterthwaite's method [
1117
      lmerModLmerTest]
1118
      Formula: lrr.rich ~ 0 + trt + (1 | site code)
1119
         Data: lrr
1120
1121
      REML criterion at convergence: -84.6
1122
1123
      Scaled residuals:
1124
         Min 10 Median
                              30
                                       Max
1125
      -5.1041 -0.5523 0.0265 0.5783 3.1723
1126
1127
      Random effects:
1128
     Groups Name Variance Std.Dev.
1129
      site code (Intercept) 0.00700 0.08367
1130
      Residual 0.04722 0.21731
1131
     Number of obs: 777, groups: site code, 34
1132
1133
      Fixed effects:
1134
             Estimate Std. Error
                                        df t value Pr(>|t|)
1135
             -0.036639 0.025203 164.432341 -1.454
                                                       0.147925
     trtK
1136
            -0.093661 0.025203 164.432341 -3.716
                                                       0.000277 ***
     trtN
1137
     trtNK -0.093698 0.025203 164.432341 -3.718
                                                       0.000275 ***
     trtNP -0.121018 0.025203 164.432341 -4.802 0.000003516801 ***
1138
1139
     trtNPK -0.169700 0.025203 164.432341 -6.733 0.00000000264 ***
1140
     trtP -0.009294 0.025203 164.432341 -0.369 0.712793
1141
      trtPK -0.067037 0.025203 164.432341 -2.660
                                                       0.008591 **
1142
      Signif. codes: 0 ,Äò***,Äô 0.001 ,Äò**,Äô 0.01 ,Äò*,Äô 0.05 ,Äò.,Äô 0.1
1143
1144
      ,Äò ,Äô 1
1145
1146
     Correlation of Fixed Effects:
1147
            trtK trtN trtNK trtNP trtNPK trtP
1148
            0.330
      trtN
1149
      trtNK 0.330 0.330
1150
     trtNP 0.330 0.330 0.330
1151
     trtNPK 0.330 0.330 0.330 0.330
1152
     trtP 0.330 0.330 0.330 0.330 0.330
1153
     trtPK 0.330 0.330 0.330 0.330 0.330 0.330
1154
```

Table S11. Model summary for nutrient treatment effects on mean species richness

1155 **Table S12.** Model summary for nutrient treatment effects on detrended species synchrony

1156 Linear mixed model fit by REML. t-tests use Satterthwaite's method [1157 lmerModLmerTest] 1158 Formula: lrr.sync ~ 0 + trt + (1 | site code) 1159 Data: lrr 1160 REML criterion at convergence: 2265 1161 1162 1163 Scaled residuals: 1164 Min 10 Median 30 Max 1165 -4.2103 -0.6279 0.0262 0.6445 2.8514 1166 1167 Random effects: 1168 Groups Name Variance Std.Dev. 1169 site_code (Intercept) 0.2244 0.4737 1170 Residual 0.9839 0.9919 1171 Number of obs: 777, groups: site code, 34 1172 1173 Fixed effects: 1174 Estimate Std. Error df t value Pr(>|t|)1175 trtK -0.03914 0.12472 119.96988 -0.314 0.7542 1176 trtN 0.08092 0.12472 119.96988 0.649 0.5177 1177 trtNK -0.23001 0.12472 119.96988 -1.844 0.0676 . trtNP-0.119160.12472119.96988-0.955trtNPK-0.145910.12472119.96988-1.170trtP-0.119470.12472119.96988-0.958 1178 0.3413 1179 0.2444 1180 0.3400 1181 trtPK -0.09823 0.12472 119.96988 -0.788 0.4325 1182 ____ 1183 Signif. codes: 0 ,Äò***,Äô 0.001 ,Äò**,Äô 0.01 ,Äò*,Äô 0.05 ,Äò.,Äô 0.1 1184 ,Äò ,Äô 1 1185 1186 Correlation of Fixed Effects: 1187 trtK trtN trtNK trtNP trtNPK trtP 1188 trtN 0.430 1189 trtNK 0.430 0.430 1190 trtNP 0.430 0.430 0.430 1191 trtNPK 0.430 0.430 0.430 0.430 1192 trtP 0.430 0.430 0.430 0.430 0.430 1193 trtPK 0.430 0.430 0.430 0.430 0.430 0.43 1194

Name	Site(s) used in analysis	Developed /framed research question(s)	Analyzed data	Contributed to data analyses	Wrote the paper	Contributed to paper writing	Site co- ordinator	Nutrient Network co- ordinator	Site acknowledgements
Oliver Carroll		х	х		х				
Evan Batzer				х		х			
Siddharth Bharath				х		х			
Elizabeth T. Borer	cdcr.us, bnch.us, look.us,					x	х	х	
	hopl.us, mcla.us, sier,us								
Sofia Campana						х			
Ellen H. Esch				x		х			
Yann Hautier	frue.ch			х		х	х		
Timothy Ohlert	sevi.us			х		х	х		
Eric W. Seabloom	cdcr.us, bnch.us, look.us, hopl.us, mcla.us, sier,us			x		x	x	х	
Peter B. Adler	shps.us					х	х		
Jonathan D. Bakker	smith.us					х	х		
Lori Biederman	cbgb.us					х	х		
Miguel N. Bugalho	comp.pt					x	x		MNB was funded by FCT contracts DL 57/2016/ CP1382/CT0030, UID/BIA/ 50027/2013 and POCI-01-0145- FEDER
Maria Caldeira	comp.pt					x	x		Funding for CEF (UID/AGR /00239/2019) and Companhia das Lezirias for site access
Qingqing Chen						х			
Kendi Davies	bldr.us					х	х		
Philip A. Fay	temple.us					х	х		
Johannes M. H. Knopps	cdrp.us					х	Х		
Kimberly Komatsu	konz.us, saline.us					х	х		
Jason Martina	temple.us					х	х		
Kevin S. McCann				х		х			
Joslin L. Moore	bogong.au					х	х		
John W. Morgan	bogong.au, kiny.au					х	х		

Table S13. Summary of author contributions and site-level acknowledgments. Site names match those in Table S1.

Table S13. Continued.

Name	Site(s) used in analysis	Developed /framed research question(s)	Analyzed data	Contributed to data analyses	Wrote the paper	Contributed to paper writing	Site co- ordinator	Nutrient Network co- ordinator	Site acknowledgements
Taofeek O. Muraina						х			
Brooke Osborne						x			
Anita C. Risch	valm.ch					х	х		
Carly Stevens	lancaster.uk					х	х		
Peter A. Wilfhart						х		х	
Laura Yahdjian						х	х		
Andrew S. MacDougall	cowi.ca	х				х	х		

- **Table S14.** Principal investigators of sites from which data were used in this analysis, but who are
- 1199 not authors. Names ordered by site. Data they provided was critical to this work.

Site Pl	Site code
Brett Melbourne	bldr.us
Jennifer Firn	burrawan.au
W. Stanley Harpole, Kirsten Hofmockel,	cbgb.us
Lauren Sullivan	
W. Stanley Harpole	cdcr.us
Elsa Cleland	elliot.us
Andy Hector	frue.ch
Rebecca McCulley, Jim Nelson	hall.us
W. Stanley Harpole	hopl.us
Mahesh Sankaran	kibber.in
Marc Cadotte, Arthur Weiss	koffler.ca
Melinda Smith	konz.us
Juan Alberti, Pedro Daleo	marc.ar
W. Stanley Harpole	mcla.us
Suzanne Prober	mtca.au
Melinda Smith	saline.us
Carka D'Antonio, W. Stanley Harpole	sedg.us
Scott Collins, Laura Ladwig	sevi.us
Dana Blumenthal, Cynthia Brown, Julia	sgs.us
Klein, Alan Knapp	
W. Stanley Harpole	sier.us
Janneke Hille Ris Lambers	smith.us
Rebecca McCulley, Jim Nelson	spin.us
Xiaohui Feng, Andrew Leakey	trel.us
Kevin Kirkman, Michelle Tedder	ukul.za
Martin Schuetz	valm.ch