

1 **TITLE:** Herbivory and eutrophication modulate grassland plant nutrient responses across a
2 global climatic gradient

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59 revisions.

60 **DATA ACCESSIBILITY STATEMENT:** Should the manuscript be accepted, the data
61 supporting the results will be archived in an appropriate public repository such as Dryad or
62 Figshare and the data DOI will be included at the end of the article.

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66

67 **ABSTRACT**

68 Plant stoichiometry, a key regulator of ecosystem functioning, is being altered by human
69 activities. We measured plant elemental responses (C, N, P and K) to eutrophication and
70 vertebrate herbivore removal at eighteen sites on six continents. Nutrient pools diverged strongly
71 in response to herbivore removal when fertilized; response ratios were highest in fenced,
72 ungrazed plots at low rainfall, whereas herbivore grazing dampened the community nutrient
73 responses to eutrophication. Across sites, climate and atmospheric N deposition emerged as
74 strong predictors of plot-level nutrients, mediated by biomass and plant chemistry. Within sites,
75 fertilization increased nutrient pools, but results were contingent on soil fertility and the
76 proportion of the biomass that was grass relative to other functional types. Our study highlights
77 (i) the role of climate in determining plant nutrient concentrations through effects on plant
78 biomass, (ii) that eutrophication affects grassland nutrient pools via both soil and atmospheric
79 pathways and (iii) interactions among soils, herbivores and eutrophication drive plant nutrient
80 responses at small scales, especially at water-limited sites.

81

82 **INTRODUCTION**

83 The relative concentration of elements in living tissues, i.e. stoichiometry, is a fundamental
84 organismal property regulating processes from cell metabolism to growth and reproduction

85 (Sterner & Elser 2002). In plants, nutrient concentrations influence primary production,
86 herbivore consumption and decomposition, thus dictating ecosystem energy flow (Elser et al.
87 2010). Macronutrients, such as N and P, play essential roles in cellular and metabolic processes –
88 N is a major constituent of photosynthetic enzymes (i.e., RuBisCO) and P is in high demand by
89 ribosomal RNA during growth and development (Elser et al. 2003). Consequently, N and P are
90 widely acknowledged as the elements that limit primary productivity (Gusewell 2004). However,
91 recent studies suggest that less well-studied elements, such as K, may also limit or co-limit
92 global plant productivity (*e.g.*, Fay et al. 2015).

93 Across terrestrial ecosystems, plant nutrient concentrations vary predictably with latitude, with
94 %N and %P decreasing towards the tropics (Reich & Oleksyn 2004, Borer et al. 2013). Nutrient
95 supply rates can directly change plant tissue chemistry (Bracken et al. 2014) and one hypothesis
96 to explain the latitudinal gradient in plant tissue chemistry is based on changing nutrient supply
97 rates with latitude. Specifically, the ‘nutrient supply hypothesis’ suggests that higher
98 temperatures near the tropics promote greater carbon fixation per unit enzyme thus diluting N,
99 while tropical soils are highly weathered thus limiting P availability (Reich & Oleksyn 2004).
100 And indeed recent theory has attempted to link temperature to plant stoichiometry across global
101 gradients (*e.g.*, Kerkoff et al. 2005, Allen & Gillooly 2009). However, the convergence in tissue
102 %N and %P across global latitudinal gradients and similar scaling of %N and %P (i.e., the slope
103 of the $\log N \sim \log P$ regression) across freshwater, marine and terrestrial ecosystems, leads to the
104 alternative hypothesis that growing season length or solar radiation reaching earth’s surface
105 (insolation) is responsible for global variation in producer nutrient concentrations (Borer et al.
106 2013).

107 Resolving the mechanisms which control plant stoichiometry within and across ecosystems has
108 important implications for understanding food web structure. Across ecosystems, producer tissue
109 %N and %P are strong predictors of the ratio of herbivore to producer (H:P) biomass and
110 herbivore consumption rates (Cebrian et al. 2009, Hillebrand et al. 2009). On the other hand,
111 herbivores alter plant stoichiometry directly by changing nutrient concentrations in re-growing
112 tissues, or indirectly, by influencing plant growth rates, resource supply, or species composition
113 (Ruess & McNaughton 1987, Hobbie 1992, Bardgett & Wardle 2003, Piñeiro et al. 2010, Cherif
114 & Loreau 2013). Herbivore influences on nutrient cycling and plant nutrient concentrations is of
115 elevated significance because humans are reducing native, large-bodied herbivore populations
116 throughout the world's grassland and savannas (*e.g.*, Craigie et al. 2010, Cebellos et al. 2015,
117 Ripple et al. 2015).

118 Another way that humans are altering earth's biogeochemical cycles is through intensified
119 agricultural practices and atmospheric nutrient deposition (*e.g.*, Vitousek et al. 1997, Stevens et
120 al. 2004). Global anthropogenic sources of N applied in fertilizers is $\sim 77.4 \pm 4.6 \text{ Tg year}^{-1}$
121 (Potter et al. 2010, Table 2), while atmospheric N deposition, derived from local fertilizer and
122 industrial sources, has nearly doubled from pre-industrial levels (from $\sim 22 \text{ Tg N yr}^{-1}$ to $\sim 39 \text{ Tg}$
123 N yr^{-1}) and is projected to double again by 2100 (IPCC 2000; Krishnamurthy et al 2007).

124 Anthropogenic nutrient inputs impact plant communities by altering plant growth rates, tissue
125 stoichiometry, rates of herbivory and community composition (*e.g.*, Pardo et al. 2011). In
126 grasslands and savannas, which cover >25% of the terrestrial biosphere (Scholes & Archer 1997,
127 Asner et al 2004), eutrophication and herbivore loss are occurring concomitantly, with important,
128 and potentially interacting, consequences for ecosystem function and biodiversity (Borer et al.
129 2014a, Hautier et al. 2014).

130 One goal of our work was to analyze the strength of local-scale disturbances, such as
131 eutrophication and the loss of large herbivores, within the context of broad-scale climate factors.
132 Studies investigating total nutrient stocks have largely focused on C and N and typically within
133 single ecosystems (*e.g.*, Schuman et al. 1999, Green & Detling 2000), with few studies seeking
134 controls across global extents (but see Wang et al. 2010). Consequently, a specific goal of our
135 study was to compare the contribution of global climate factors, including temperature,
136 precipitation, solar radiation and N deposition (*e.g.*, Stevens et al. 2015), to variation in
137 herbivory and nutrient supply on nutrient concentrations at the plant- (*e.g.*, nutrient
138 concentrations, g g^{-1} tissue) and community-level (*e.g.*, on an areal basis, g m^{-2}). As herbivore
139 and fertilization effects may depend on climate and nutrient supply, we also tested for statistical
140 interactions among eutrophication, herbivory, climate and variation in soil fertility across a
141 global range of sites.

142 In our first analysis, we experimentally manipulated nutrient supply (NPK fertilization) and
143 herbivory (fencing) and used an analysis of covariance to ask how the relative quantity of
144 nutrients (C, N, P and K) in the standing biomass (measured in g m^{-2} , hereafter ‘total’ nutrients)
145 varied across a global climate gradient (model one). Subsequently, we analyzed the network of
146 direct and indirect effects of climate, herbivores and resource supply on total plant nutrient
147 responses (the sum of N, P and K) at multiple spatial scales by combining site- and plot-level
148 predictors in a multi-level structural equation model (SEM, model two). For model one, we
149 predicted an interaction between eutrophication and herbivory across sites: specifically, that the
150 relative effects of eutrophication would increase with rainfall, but the response would be
151 amplified by increased abundance of nutrient-rich palatable plants at higher rainfall sites when
152 herbivores were removed. This prediction results from work demonstrating herbivore effects on

153 plant species turnover are greater at mesic compared to arid sites (*e.g.*, Chase et al. 2000). On the
154 other hand, if nutrient availability across sites results largely from herbivore effects on biomass
155 instead of plant turnover, then herbivore loss may lead to greater impacts of eutrophication at
156 arid sites, where herbivore effects on biomass are the largest (*e.g.*, Chase et al. 2000).

157 For model two (the multi-level SEM), we decomposed the response of total plant nutrients into
158 those due to changes in tissue nutrient concentration and those driven by changes in plant
159 biomass. We expected concentrations of N, P and K to decrease with temperature across sites, as
160 reported elsewhere (*e.g.*, Reich and Oleksyn 2004; Borer et al. 2013). Within sites we predicted
161 that fertilization and herbivory would have strong direct effects on nutrient responses through
162 their effects on plant chemistry and plant biomass, but that these effects would depend on
163 background resource supply, *i.e.*, the largest responses in low soil nutrient plots. Finally, in terms
164 of their strength in controlling variation in total plant nutrient pools, we predicted that plot-level
165 (*e.g.*, herbivores, eutrophication and resource supply) would be comparable in magnitude to
166 broad-scale climatic drivers (*e.g.*, rainfall, temperature, insolation and N-deposition).

167

168 **METHODS**

169 *Site location and study design*

170 Our study was conducted at 18 sites in Africa, Asia, Australia, Europe, North and South America
171 (Fig. 1) as part of the Nutrient Network (NutNet; Borer et al. 2014b). Sites represented a range of
172 habitat types and spanned a gradient of mean annual precipitation (MAP) from 305 to 2315 mm
173 yr⁻¹ (Table S1 in Supporting Information), but all were located in grass-dominated plant
174 communities. At each site, fully factorial combinations of nutrient addition and fencing were

175 randomly assigned to 25-m² plots arranged in 1 - 5 blocks within sites from 2007 - 2012 (Table
176 S1). Within each 25-m² plot, randomly chosen 1-m² quadrats were selected in one of four sectors
177 for continuous data collection. In nutrient addition plots, NPK was added annually at the onset of
178 the growing season as a combination of nutrients at the following rates: slow release urea (10 g
179 N m⁻² yr⁻¹), triple super phosphate (10 g P m⁻² yr⁻¹), and potassium sulfate (10 g K m⁻² yr⁻¹). In
180 year one only, 100 g m⁻² of a micronutrient mix (Fe, S, Mg, Mn, Cu, Zn, B and Mo) was applied
181 in the nutrient addition plots. In herbivore exclusion plots, fences were constructed of heavy
182 gauge wire to heights of 120 - 180 cm designed to exclude mammalian herbivores > 50 g. To
183 exclude small mammals, most sites included a 1 cm wire mesh secured at the base of each fence
184 to a height of 90 cm. NutNet sites include all combinations of N, P and K nutrient addition in the
185 absence of fences (*e.g.*, Fay et al. 2015), but those treatments are ignored given our emphasis on
186 herbivore-nutrient interactions.

187 *Sample collection and nutrient analyses*

188 Plant biomass and tissue concentration of key elements (C, N, P, and K) (Elser et al. 2003; Fay
189 et. al 2015) were sampled 3 - 5 years after the establishment of the experiment at 16 sites and
190 after either 1 or 2 years at the remaining two sites (Table S1). Aboveground plant biomass was
191 clipped to ground-level in 2 replicate 10 x 100 cm strips, sorted by functional type (grass, forb,
192 legume and woody plants), dried and weighed to the nearest 0.1 g. Samples were transported to
193 Wake Forest University where they were ground in an UDY belt-drive sample mill, dried at 65
194 °C for 48 h and analyzed, by functional type, on a Bruker near infrared spectrophotometer
195 (NIRS) (Bruker Optics, Ettlingen, Germany). Reflectance data from each ground sample were
196 averaged from triplicate measurements between wavelengths of 781 – 2778 nm (12,800 – 3,600

197 cm^{-1}) at 16 nanometer resolution with the rotating cup method, except for samples < 3 g, in
198 which case samples were analyzed in a stationary vial (~19 % of samples).

199 Spectral data were used to identify a subset of samples (20 % of the total sample number),
200 stratified by site, for wet chemical analysis, i.e., the ‘known’ calibration samples (Appendix S1).
201 Samples were analyzed for total C, N, P, and K at either North Carolina State University or
202 Kansas State University using standard analytical methods. The resulting dataset was further
203 subdivided into a calibration (model development) and validation (test set; 10%) subset using an
204 algorithmic experimental design approach (Appendix S1).

205 Soils were collected 2 - 4 years post treatment (mean = 3.4 ± 0.8 years; Table S1) in each subplot
206 to 10 cm (approximately 250 g soil), bagged, air-dried and sent to the University of Nebraska for
207 archiving and total %C and %N analysis via dry combustion gas chromatography (Dumas
208 method, COSTECH ESC 4010 Element Analyzer). As our interests were in soil fertility and its
209 interaction with treatments, we focused on soil %N, as it is a widely acknowledged indicator of
210 soil fertility and a major plant-limiting nutrient (*e.g.*, Elser et al. 2007). Soils were unavailable
211 from three sites (Table S1) and soil %N was imputed for these sites because of the relatively
212 strong relationship between soil N and major climate factors (Appendix S1). Further details on
213 sampling methodology are at http://www.nutnet.org/exp_protocol.

214 *Data analysis*

215 *Effect of fertilization on total plot nutrients in grazed vs. ungrazed grassland (model one)* - In
216 our first model, we analyzed the effects of eutrophication and herbivore removal on total plot
217 nutrients within the context of global environmental variation. To analyze plot-level responses to
218 herbivores and eutrophication, we asked if the relative responses of total plot nutrient content to

219 fertilization depended on climate or soil fertility, and if the presence of herbivores altered the
220 relationship. We focused on three climate factors with strong conceptual and empirical links to
221 plant stoichiometry: temperature, rainfall, and solar insolation (Table S1). For each site, we
222 extracted mean annual temperature (MAT, °C), variable BIO1, and mean annual precipitation
223 (MAP, mm yr⁻¹), BIO12, from BioClim (Hijmans et al. 2005). For solar insolation (INS, kWh m⁻²
224 day⁻¹), we extracted average annual data (1983-2005) from the NASA Surface meteorology and
225 Solar Energy database (<http://eosweb.larc.nasa.gov/sse/>). Soil fertility at the plot-level was
226 represented in the models by SOIL %N as described above.

227 For this analysis, all plots subjected to fencing (FENCE; fenced versus control) and fertilizer
228 addition treatments (NPK; fertilized versus control) were included (2 fencing levels x 2 NPK
229 addition levels = 4 treatment combinations per block). Plot-level estimates of each nutrient (C,
230 N, P and K) were obtained by summing, for all functional types in a plot, the product of their
231 tissue nutrient concentration (% dry weight) and biomass in g m⁻², yielding the total nutrient
232 content in aboveground biomass (g m⁻²) for each element. We then quantified the relative effects
233 of nutrient addition by calculating log response ratios (LRR) within blocks at each site:

234 $\log\left(\frac{\text{total nutrient content in NPK addition}}{\text{total nutrient content in control}}\right)$. LRR for each block and site were plotted against INS,

235 MAP, MAT and soil %N for both fenced and unfenced treatments. An analysis of covariance
236 was used to determine if the slopes of the LRR ~ environmental predictors were different for
237 levels of FENCE.

238 To identify the best model, we followed a model selection procedure on the basis Akaike's
239 Information Criteria modified for small sample sizes (AICc; Burnham and Anderson 2002).
240 Candidate models included all main effects plus environment by treatment interactions. Models
241 were selected as best fits to the data when ΔAICc values were < 2 below that of other models.

242 For models within a 1 Δ AIC unit of each other, the model with the fewest parameters was
243 selected as the final, accepted model. After identifying the most suitable model (see below), we
244 tested for interactions among NPK addition, fencing and environmental predictors using type III
245 sums of squares using the *Anova* command in R-package ‘car’.

246 Structural equation model of total plot NPK (model two) - In our second analysis, we used
247 structural equation modeling (SEM) to quantify system-level influences of climate, soil fertility,
248 herbivory and eutrophication on total plot nutrients. For this analysis, rather than analyze each
249 nutrient separately, total N, P, and K were summed on an areal basis (g m^{-2}) to create a single
250 nutrient variable that was modeled as the response. As the total plot nutrient content is a product
251 of multiple direct and indirect sources, our *a priori* model was driven by variation in three
252 sources: (1) direct effects due to plant chemistry (*i.e.*, $\text{g nutrient g plant}^{-1}$), (2) direct effects due
253 to plant community biomass (g m^{-2}) or (3) indirect effects due to variation in the abundance of
254 functional types (*i.e.*, grass vs. forb) among sites (Figure S1 and Table S2). Due to the
255 dominance of grasses across the sites and their important functional role, percent grass biomass
256 (‘% grass’) was included to account for functional type turnover among sites. The %grass
257 variable was logit transformed prior to analyses to meet model assumptions. External predictors
258 were MAT, MAP, INS, atmospheric N deposition, soil % N and the two treatment variables,
259 NPK and FENCE, as discrete binomial predictors. In addition, we included a ‘grazing index’ that
260 accounted for site-level variation in herbivore abundance and diversity (Appendix S1). Due to
261 the hierarchical nature of the data (*i.e.*, some predictors existed only at the site level while others
262 existed for individual plots; Figure S1) we analyzed each response variable in a piecewise
263 fashion using a multi-level approach (*e.g.*, Gelman & Hill 2007; Appendix S1). Note that the
264 final response variable in the SEM, total standing NPK, is a mathematical product of the quantity

265 of plant material in a plot and the nutrient concentration in plant tissue. For this reason,
266 standardized path coefficients connecting plot biomass and plant chemistry to total standing NPK
267 were computed analytically rather than estimated. These computed parameters represent the
268 contributions to variations in total standing NPK derived from variation in component variables
269 (Appendix S1).

270

271 **RESULTS**

272 *Effects of eutrophication in unfenced (grazed) vs. fenced (ungrazed) grassland (model one)*

273 The ANCOVA analysis (*i.e.*, LRRs) demonstrated that element responses to eutrophication
274 across a global gradient in rainfall depended on the experimental removal of herbivores. For each
275 of the elements analyzed (C, N, P and K), the top model identified by AICc included an
276 interaction between MAP and FENCE on element LRR under fertilization (Table S3). All
277 elements showed a consistently strong negative relationship between the LRR and MAP inside
278 ungrazed, fenced plots, meaning that the strongest effects of increasing nutrient supply were at
279 arid sites in the absence of herbivores (Fig. 2). The presence of herbivores counteracted the
280 strong effects of NPK addition at dry sites, demonstrated by the flat relationship between MAP
281 and LRR in grazed plots across a gradient of MAP ($P > 0.1$ for hypotheses that slopes and
282 intercepts were non-zero in a linear models for all elements in the ungrazed treatments). No other
283 model was similar in its fit with LRR across sites (Table S3) and the final coefficient of
284 variations (R^2) for the models were between 26% (for plot P) and 32% (for plot N).

285 *SE model results of total plot NPK (model two)*

286 For the SEM, our initial overall hypothesis was that herbivore removal and eutrophication would
287 alter total grassland nutrients by influencing plant chemistry and plant composition, and that their
288 effects would be similar in magnitude. The final SE model had coefficients of determination (R^2)
289 of 0.33 for percent grass, 0.49 for total plot biomass and 0.67 for total plant NPK (Fig. 3). Here,
290 we present standardized path coefficients for the final SE model (both standardized and
291 unstandardized coefficients are presented in Tables S4 and S5). The model results support the
292 interpretation that there are strong direct effects of climate variables, especially MAT and INS,
293 on %grass and plot biomass, and somewhat weaker influences of climate on plant chemistry
294 (plant NPK). MAT influences on %grass and biomass were positive (1.06 and 1.08), but plant
295 NPK decreased with MAT (-0.63). The only direct influence of MAP in the model was a positive
296 effect on plant NPK (0.50). INS had relatively strong negative influences on both %grass (-0.88)
297 and plot biomass (-1.0). After accounting for climate effects, N deposition increased plot
298 biomass (0.29) and decreased plant NPK (-0.37), likely because of growth-induced nutrient
299 dilution in plant tissues.

300 While fencing was not significant in the final SE model, the grazer index was positively related
301 to %grass (0.42) and plant NPK (0.22) at the site-level. At the plot-level, there was a relatively
302 weak response of plant NPK to an interaction between NPK fertilizer and soil %N (Fig. 3) and a
303 somewhat stronger response of plot biomass to an interaction between NPK fertilization and
304 %grass (0.19; Fig. 3). For plant NPK, the interaction arose from a positive response of plant
305 NPK to the soil fertility gradient (soil %N) in unfertilized plots and a negative response in
306 fertilized plots (Fig. 4). For the plot biomass, the interaction arose from a positive relationship
307 between %grass in plots and total biomass in the absence of NPK fertilization and a negative
308 relationship for plots fertilized with NPK. However, inspection of the relationship demonstrates

309 that the interaction is driven by a large biomass response at low %grass in fertilized plots and a
310 relatively stable response of high %grass plots to fertilization (Fig. 5).

311 After computing standardized coefficients, plot biomass had 2.7 times the influence on the
312 variance in total standing NPK compared to plant NPK (0.93 vs. 0.35). In terms of total effects
313 on total standing NPK (i.e., direct + indirect effects), INS had the strongest effect (-0.91), which
314 was mediated by a strong negative influence on plot biomass and a strong positive relationship
315 between biomass and standing NPK (Table S5). MAT had the strongest positive effect on plot
316 standing NPK content (0.72), which was mediated by its strong positive association with plot
317 biomass (Table S5). MAP had a weaker positive effect on total standing NPK (0.18), which was
318 mediated by its positive effects on plant NPK.

319 MAT and N deposition were unique in the model in that both predictors had positive effects on
320 biomass that were offset by negative effects on plant nutrient concentrations (Table S5). For
321 example, for N deposition, the increase in total standing NPK due to greater biomass (0.27) was
322 offset by a decrease in total standing NPK due to lower plant NPK (-0.13), which dampened the
323 overall positive influence of N deposition on community nutrient pools. Herbivore abundance, as
324 measured by the grazing index, increased total standing NPK by increasing plant NPK (Fig. 3).
325 However, these were offset by a positive relationship between the grazer index on %grass, which
326 reduced plant NPK; the result was a weak overall increase in total standing NPK (0.05; Table
327 S5). Finally, the effects of nutrient additions depended on the background plant community
328 (%grass) and underlying resource availability (soil %N). However, the total strength of
329 eutrophication, which includes interactions with both %grass and soil %N, accounted for a
330 consistent positive effect on total standing NPK (0.23).

331

332 **DISCUSSION**

333 Across our global sampling of grassland sites, climate variation best explained broad-scale
334 patterns of nutrient concentrations, but these effects were modified locally by eutrophication and
335 herbivory. For example, elements showed a consistently strong negative relationship between the
336 LRR and MAP in ungrazed, fenced plots, whereas response ratios were flat and not different
337 from zero across a global precipitation gradient in the presence of herbivores (Fig. 2). These
338 results contrast with our initial predictions and suggest that herbivore effects on element standing
339 stocks are dominated by their consumptive effects on biomass, rather than their effects on plant
340 species compositional turnover, which increase with precipitation (*e.g.*, Chase et al. 2000,
341 Anderson 2008). However, our results are consistent with studies showing consumers have their
342 greatest proportional effect (*e.g.*, on productivity) in arid relative to mesic sites (Olf & Ritchie
343 1998, Chase et al. 2000).

344 Another implication of these results is that, across a global range of sites, herbivory fully
345 compensated for plot-level nutrient production after experimental eutrophication. Because plant
346 nutritional quality acts as a key regulator of decomposition and carbon storage (*e.g.*, Cebrian
347 1999), the outcome of our experiment suggests that the continued loss of large herbivores from
348 ecosystems (*e.g.*, Ripple et al. 2015) will further compound effects of anthropogenic
349 eutrophication on ecosystem processes.

350 In the final SE model, INS and MAT provided the greatest explanatory power of global variation
351 in total standing quantities of NPK in vegetation. Our findings are consistent with the hypothesis
352 that solar irradiance and temperature are major drivers of global variation in plant nutrients
353 across the earth's surface (Borer et al. 2013). Our explicit test of the solar insolation hypothesis
354 found strong support (Figure 3, Table S5), although the major influence was through negative

355 effects on biomass and %grass rather than on plant chemistry *per se*. Our results suggest that the
356 latitudinal decline in NPK observed in grassland plants (*e.g.*, Reich & Oleksyn 2004, He et al.
357 2008, Borer et al. 2013) arises from direct effects of MAT and MAP on plant nutrient
358 concentrations at the plant-scale (*i.e.*, paths from MAT and MAP to plant NPK in Fig. 3) and
359 indirect effects, mediated by biomass, at the plot-scale (*i.e.*, path from insolation and MAT to
360 plot biomass in Fig. 3).

361 As a consequence of the 2.7 times greater sensitivity of total standing NPK to plant biomass
362 compared to plant chemistry (standardized path coefficients of 0.93 compared to 0.35), the
363 factors with the largest influence on plant biomass have the greatest impact on total nutrient
364 flows in grasslands. Perhaps not unexpected on its own, this is surprising given that much of the
365 research on plant stoichiometry has focused on patterns of variation in plant-level chemistry at
366 global (*e.g.*, Craine et al. 2002, Reich & Oleksyn 2004, Borer et al. 2013), regional (*e.g.*, He et
367 al. 2006, Zhang et al. 2012), functional type (*e.g.*, Han et al. 2011) or phylogenetic (Stock &
368 Verboom 2012) scales while ignoring the consequences of variation in plant biomass for nutrient
369 availability. Even though changes in plant composition can modify nutrient content on a mass
370 basis, such as the strong effects of legumes on %N (Spehn et al. 2002), our results suggest that
371 such influences are relatively small compared to processes that influence primary production.

372 Another clear pattern that emerged from our study is that anthropogenic eutrophication has
373 complex effects on nutrient availability across environmental gradients. First, eutrophication has
374 two pathways by which it can alter plant nutrients, one atmospheric and the other by
375 anthropogenic fertilizers applied to soil. Stevens et al. (2015) showed that N deposition was a
376 strong predictor of grassland primary production, better even than fertilization. Our results
377 demonstrate both pathways have offsetting influences on total NPK availability which are

378 mediated by positive effects on plot biomass and negative effects on the concentration of NPK in
379 plants. In the case of N deposition, the offsetting paths result in a positive effect on total plot
380 NPK (Table S5), but in the case of fertilization the response is further complicated by the
381 interaction with soil nutrients (reduction of plant NPK at fertile sites).

382 While the strong link to climate is consistent with the growth rate hypothesis, it is difficult to
383 separate effects of nutrient availability insofar as temperature and moisture modify
384 decomposition, chemical weathering and other factors that drive plant nutrient availability
385 (O'Halloran et al. 2012). Studies (*e.g.*, Craine et al. 2005, Kraft et al. 2008, He et al. 2009)
386 suggest that plants do not simply reflect the nutrient availability of a site, but instead that plant
387 chemistry is determined by a complex balance of taxonomic identity, competition and resource
388 supply rates. Indeed, our SEM analysis demonstrated that when the effects of eutrophication
389 were present, the specific nature of the outcome often interacted with background soil fertility or
390 composition of the plant community. Finally, the lack of a MAP effect on plot biomass was
391 surprising, but is consistent with regional contingencies in the effects of MAP on biomass across
392 grassland sites (*e.g.*, O'Halloran et al. 2013) and the fact that across 42 NutNet sites MAP had no
393 effect on plot ANPP, whereas atmospheric deposition significantly increased site level ANPP
394 (Stevens et al. 2015).

395 *Conclusion*

396 Our study highlights the importance of global climate gradients in establishing across site
397 variation in nutrients at the plant- and plot-level. In particular, solar insolation, MAT and MAP
398 emerged as major drivers of among-site variation in grassland nutrient pools through direct
399 effects on plant chemistry and indirect effects on plant biomass. However, consistent responses
400 to eutrophication within sites was a key driver of plant nutrient responses. Moreover, herbivores

401 dampen the effects of eutrophication on nutrient standing stocks through their consumption,
402 especially at sites where water is limiting. Consequently, continued loss of herbivore diversity
403 and increased eutrophication may disproportionately increase standing nutrients in dry areas. In
404 the absence of the diversity-promoting effects of herbivores (*e.g.*, Borer et al. 2014a, Yang et al.
405 2015) arid sites may become further destabilized by nitrogen addition (Hautier et al 2014)
406 leading to impacted rates of nutrient cycling in these regions. We suggest that a full
407 understanding of nutrient dynamics and energy flow in savanna and grassland ecosystems
408 requires a hierarchical and multivariate approach to the various ecological drivers. Finally, we
409 recommend that future efforts to map the global distribution of nutrients in grassland forage
410 (*e.g.*, Wang et al. 2010) should include climate, eutrophication and herbivore distributions in
411 future models.

412

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427

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588 **Figure Legends**

589 **Figure 1.** Global map showing the distribution of the 18 NutNet sites across 5 continents
590 represented in our analysis. See Table S1 for the specific site information.

591 **Figure 2.** Effect of fertilizer addition, expressed as a log response ratio (LRR: \log [NPK
592 addition/control]) on total plot carbon (A), nitrogen (B), phosphorous (C) and potassium (D)
593 plotted against mean annual rainfall from grassland plots from 18 sites inside (closed circles,
594 solid lines) and outside (open circles dashed lines) fences. Solid and dashed lines represent
595 significantly different fencing x MAP interaction across sites (see Table S4 in Supplementary
596 Results).

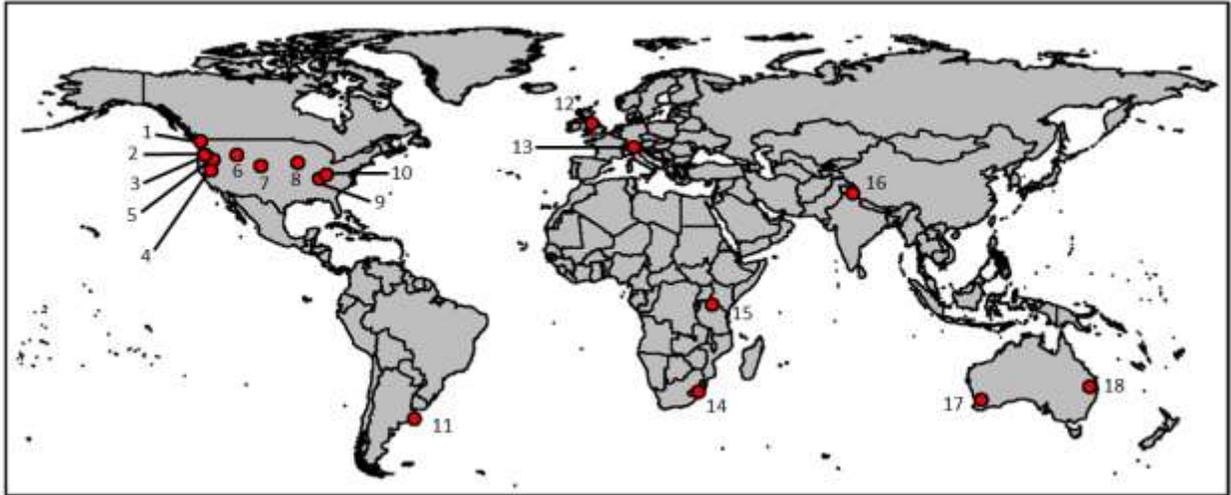
597 **Figure 3.** Final structural equation model results. The figure depicts direct and indirect effects of
598 predictors (climate [insolation, MAT and MAP], grazing index, treatment effects [fences and
599 NPK addition] and soil %N) on response variables (% grass, plot biomass [g m^{-2}], plant NPK
600 [sum of N, P and K in % dry weight] and total standing NPK [sum of N, P and K in g m^{-2}]).
601 Response variables are shown in solid and predictor variables are shown in bold-dashed boxes.
602 Predictors of a similar type are grouped within bold-dashed boxes to facilitate interpretation
603 (such as climate variables and treatments). The hierarchical structure of the model is represented
604 by thin-dashed boxed which enclose predictors into site- and plot-level groups. Interaction
605 effects are depicted with grey arrows pointing to a solid grey dot which points at the response
606 variable. Curved arrows represent correlations while straight arrows imply causal effects of one
607 variable on another with arrow widths proportional to the standardized path coefficient (shown
608 on each arrow). Numbers on arrows are standardized path strengths with arrow widths
609 proportional their values; blue arrows show positive and orange negative relationships. Bold
610 numbers associated with the response variables are total variance explained (R^2) by all

611 significant predictors. Note that total standing NPK in a plot is a mathematical product of the
612 plot biomass and plant nutrient concentration; therefore, these path coefficients (labeled with ‘*’
613 in the diagram) are computed rather than estimated and they represent the sensitivity of
614 variations in total standing NPK derived from variation in component variables. See Table S5 in
615 Supplementary Results for standardized total effects of predictors on the response variables.

616 **Figure 4.** Relationship between plot soil fertility (soil %N) and plant NPK (% dry weight) in
617 control plots (left panel) and experimentally fertilized plots (right panel) across 18 sites. Points
618 are replicate plots within sites and are labelled in order from low (orange) to high (blue)
619 precipitation; for site information and identities see Table S1. Lines show random intercepts fit
620 to individual site with the mean interaction response (either -NPK or +NPK) shown as the bold,
621 black line.

622 **Figure 5.** Relationship between logit transformed grass abundance (%) and total plot biomass (g
623 m⁻²) in control plots (left panel) and experimentally fertilized plots (right panel) across 18 sites.
624 Points are replicate plots within sites and are labelled in order from low (orange) to high (blue)
625 precipitation; for site information and identities see Table S1. Lines show random intercepts fit
626 to individual site with the mean interaction response (either -NPK or +NPK) shown as the bold,
627 black line.

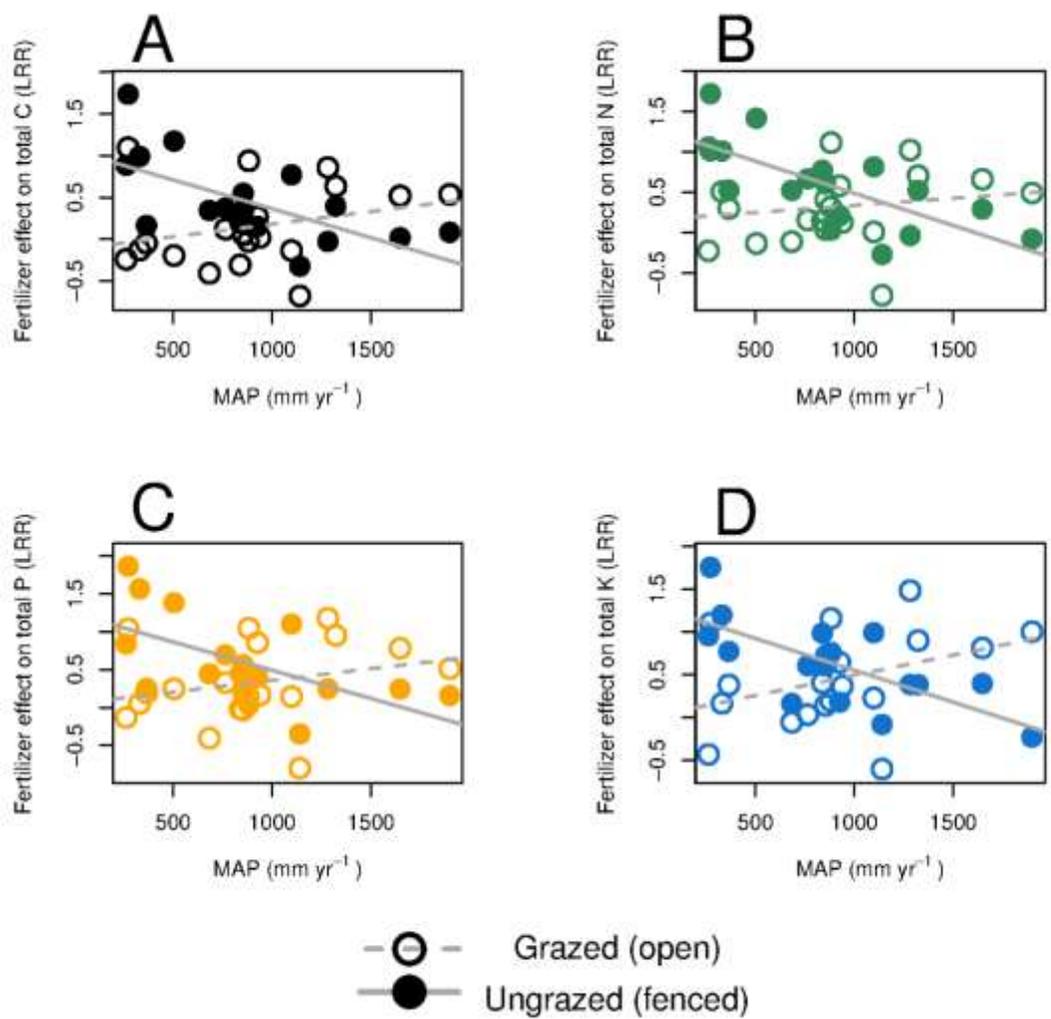
628 **Figure 1.**



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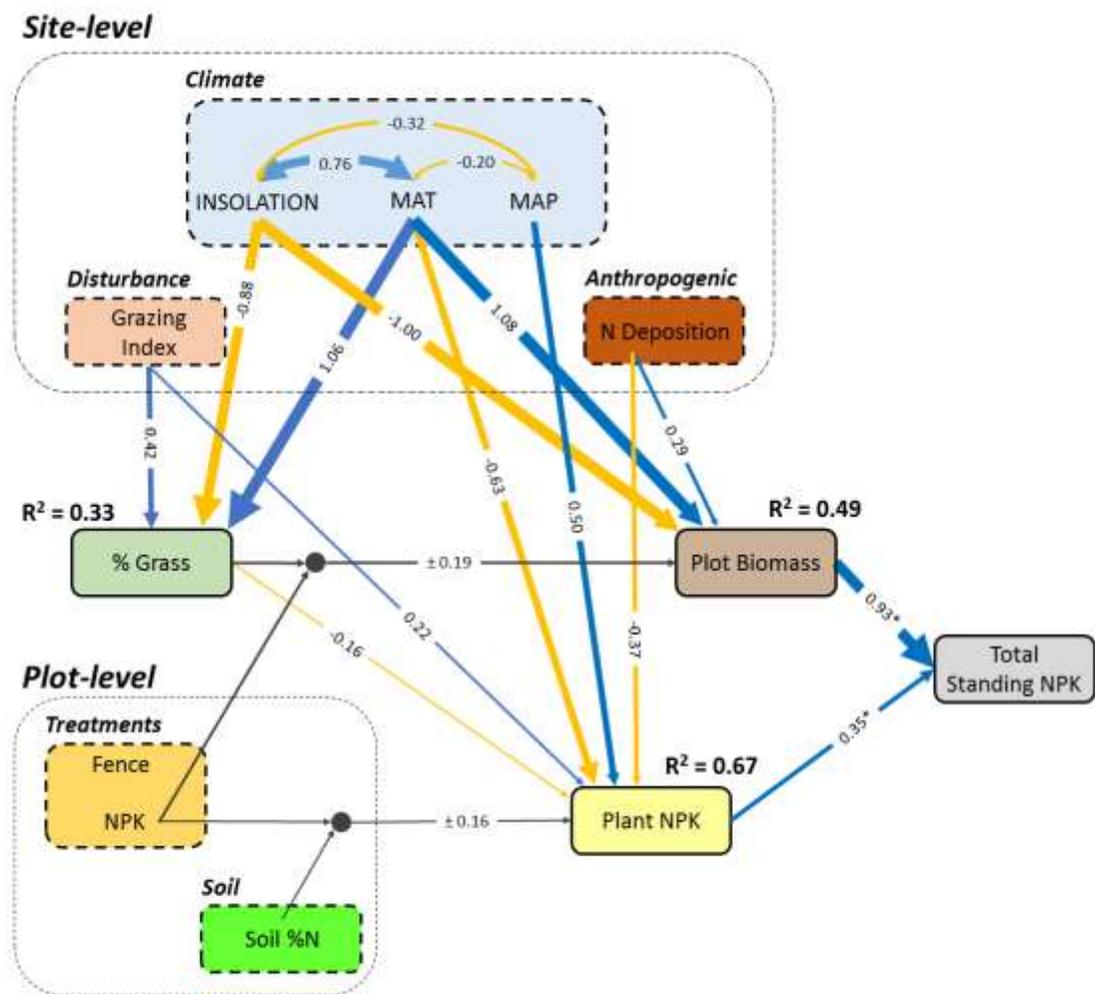
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631 **Figure 2.**



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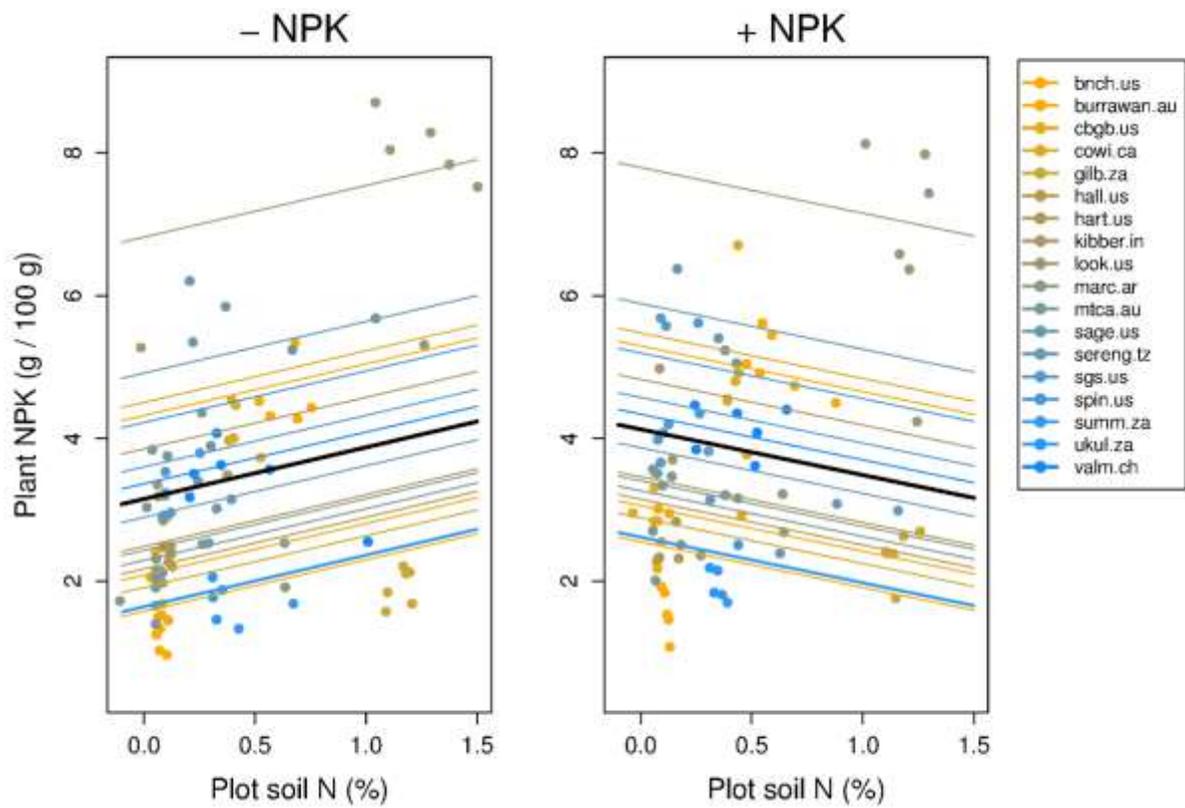
633 **Figure 3.**



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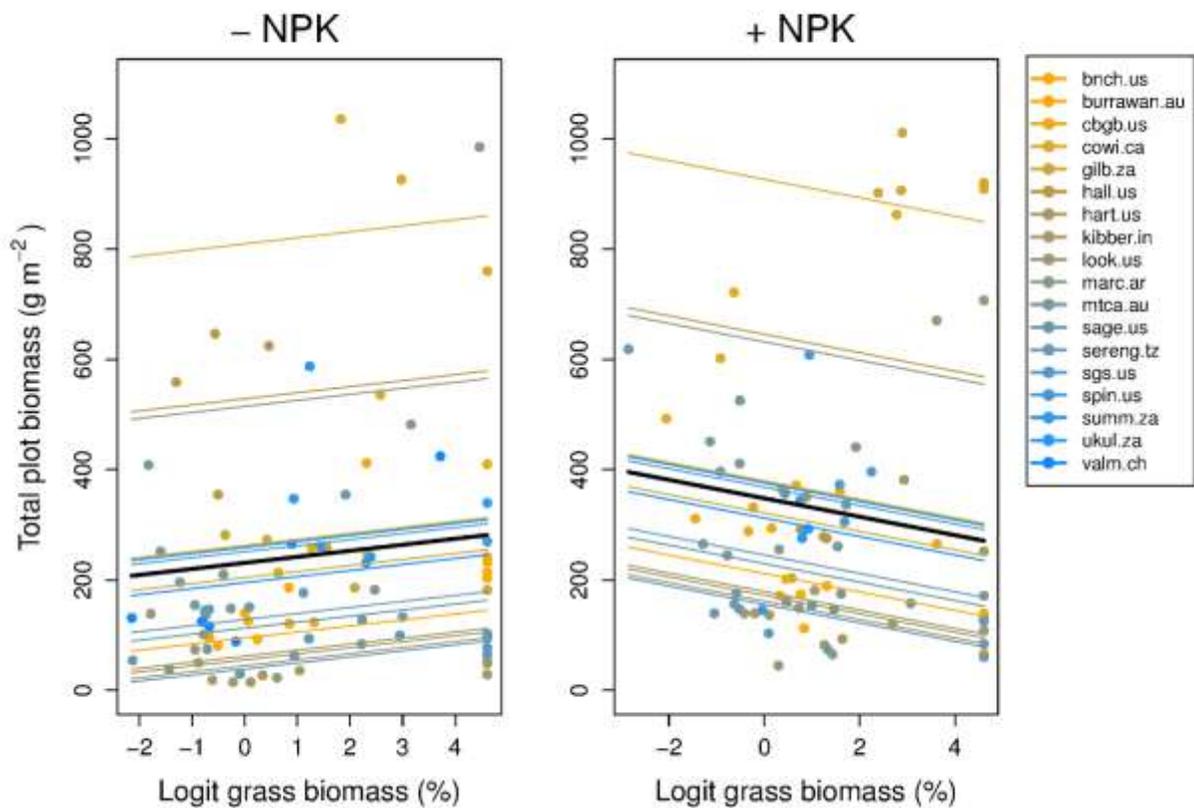
636 **Figure 4.**



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638

639 **Figure 5.**



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