

1 **Herbivore population differences rival geographic and biophysical variation in structuring**  
2 **ecosystem function**

3 Running title: Herbivore populations restructure ecosystems

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14 soil carbon; herbivory; plant-herbivore interactions; microclimate

15    **Abstract**

16    Geographic variation in ecosystem function is often attributed to differences in climate and soil  
17    properties, with biophysical constraints assumed to dictate spatial patterns in nutrient cycling,  
18    carbon storage, and plant productivity. However, biotic interactions, particularly herbivory, also  
19    vary geographically and can generate feedbacks that influence ecosystem processes. Using a  
20    replicated three-year field experiment, we tested how population-level functional differences in a  
21    widespread arthropod herbivore mediate geographic variation in ecosystem function. Structural  
22    equation modeling revealed that herbivores exerted strong direct effects on plant biomass, soil  
23    carbon, and nitrogen mineralization, often surpassing the influence of historical conditions and  
24    geographic variation in climate. Moreover, functionally distinct herbivore populations had  
25    divergent effects on nutrient cycling and plant diversity, demonstrating that population-level  
26    differences introduce novel pathways of influence on ecosystem function. These findings  
27    challenge ecosystem models that prioritize abiotic constraints and highlight the need to incorporate  
28    consumer-driven feedbacks into ecological frameworks.

29 **Introduction**

30 Geographic variation in ecosystem function has been widely attributed to spatial  
31 differences in biophysical factors, particularly climate variables such as temperature and  
32 precipitation, and legacy states such as soil properties and nutrient availability (Berg *et al.* 1993;  
33 Chapin *et al.* 2011; Reichstein *et al.* 2014; Schlesinger 2005). These factors shape broad-scale  
34 patterns in nutrient cycling, carbon storage, and primary production, leading to predictable  
35 ecosystem differences across climatic and edaphic gradients. However, ecosystem function is also  
36 influenced by biotic interactions, particularly those between herbivores and plants, which also vary  
37 geographically in response to local environmental conditions (Barley *et al.* 2021; Lynn & Fridley  
38 2019; Marczak *et al.* 2013; Massad *et al.* 2024). Herbivore-plant interactions are shaped by  
39 temperature, soil, and plant nutrient availability, generating feedback that structure nutrient cycling  
40 and ecosystem processes (Elser *et al.* 2000; Pennings *et al.* 2009; Schmitz 2017; Schmitz &  
41 Trussell 2016). Thus, while biophysical variation sets the stage for ecosystem function, spatial  
42 heterogeneity in biotic interactions is the play that may further drive regional variation in  
43 ecosystem dynamics.

44 Despite increasing recognition of biotic interactions as key drivers of ecosystem processes,  
45 the mechanisms shaping geographic variation in herbivore-plant interactions remain unresolved,  
46 particularly in the context of conflating or confounding influences from local biophysical  
47 constraints and broader geographic gradients (Lynn *et al.* 2023; Lynn & Fridley 2019; Maron *et*  
48 *al.* 2014; Massad *et al.* 2024; Pennings & Silliman 2005; Schmitz & Trussell 2016). Extensive  
49 research has shown that geographically structured soil properties influence herbivory by shaping  
50 plant defense traits (e.g., Lynn *et al.* 2023; Lynn & Fridley, 2019), and that herbivore impacts can  
51 vary depending on interactions between abiotic and biotic factors across environmental gradients

52 (Maron *et al.* 2014; Massad *et al.* 2024). Moreover, herbivory often correlates more closely with  
53 local resource availability and biophysical constraints than with latitude alone, demonstrating the  
54 importance of small-scale environmental heterogeneity (Pennings & Silliman 2005, Baker *et al.*  
55 *in review*). While this work has clarified how bottom-up drivers such as soil fertility and plant  
56 defenses shape geographic patterns in herbivory, comparatively less is known about how  
57 herbivores, in turn, generate top-down effects that influence ecosystem function across space. In  
58 particular, little attention has been given to whether spatial variation in herbivore traits—driven by  
59 local environmental conditions—can create systematic differences in how herbivores affect  
60 nutrient cycling and plant community composition. Addressing this gap requires integrating  
61 biophysical context with an understanding of trait variation within species.

62 Trait variation within a species can strongly influence individual-level impacts on  
63 ecosystem function (Govaert *et al.* 2024; Raffard *et al.* 2019; Des Roches *et al.* 2018, 2021; but  
64 see Pichon *et al.* 2022), yet this variation is often overlooked in studies of geographic ecosystem  
65 dynamics. If herbivore populations differ in functional traits, such as foraging behavior (Joern *et*  
66 *al.* 2012), nutrient excretion, or stress physiology (Rosenblatt *et al.* 2019; Sommer *et al.* 2023),  
67 then their ecosystem impacts may diverge in ways not readily predicted by climate or edaphic  
68 properties alone. Variation in trait expression can arise through local adaptation, where populations  
69 evolve traits suited to persistent local environmental conditions, or through phenotypic plasticity,  
70 where individuals modify traits in response to locally varying environmental conditions (Bradshaw  
71 & Holzapfel 2006; Reed *et al.* 2010; Sommer *et al.* 2024). This dynamic may have cascading  
72 effects on the extent to which local herbivore populations interact with plant communities and  
73 mediate ecosystem functions such as nutrient cycling. For instance, herbivores in resource-limited  
74 environments may exhibit compensatory feeding strategies, such as increasing their consumption

75 rates of carbon-rich plants, leading to greater production of frass and litter inputs with altered  
76 stoichiometry, which can accelerate decomposition and mineralization rates (Raubenheimer *et al.*  
77 2009; Sitters *et al.* 2020). By contrast, herbivores in resource-rich environments may primarily  
78 alter plant competitive dynamics, shifting community composition without strongly changing litter  
79 quality or quantity, and thereby exerting weaker influence on soil nutrient processes. These  
80 divergent strategies align with broader ecological expectations of how populations regulate energy  
81 and nutrient demands under different environmental constraints. Thus, failing to account for  
82 herbivore population-level functional differences could obscure key drivers of geographic  
83 variation in ecosystem function and resilience to environmental change.

84 We report on findings from a three-year geographically replicated experiment in which  
85 local populations of a dominant herbivore species were reciprocally transplanted across five sites  
86 to quantify how population-level trait differences mediate variation in ecosystem function across  
87 their geographic range in the New England region of the eastern USA. The species (*Melanoplus*  
88 *femurrrubrum*) is a widespread grasshopper in which local populations differ in their expression of  
89 plasticity in physiological and behavioral traits (Baker *et al.* *in review*; Parsons & Joern 2014;  
90 Rosenblatt *et al.* 2019; Sommer *et al.* 2025). We applied structural equation modeling (SEM) to  
91 quantify the relative contributions of among-population variation, local climate, and historical  
92 legacies as drivers of key ecosystem variables and functions, including plant biomass, soil  
93 nutrients, and nitrogen mineralization. Grasshopper populations originating from warmer sites,  
94 characterized by higher daily mean maximum temperatures during the growing season, drove  
95 different ecosystem outcomes than populations from cooler sites (Baker *et al.* *in review*). These  
96 differences were linked to variation in behavioral and physiological trait plasticity expressed under  
97 environmental stressors. Overall, our findings demonstrate that within-species variation in

98 herbivore traits can manifest spatially to play an important role in shaping geographic variation in  
99 ecosystem function.

100 **Methods**

101 *Natural history background*

102 Our study was conducted within a 17,000 km<sup>2</sup> area of the New England region of the  
103 northeastern USA. Within this region, old field ecosystems are localized openings in a largely  
104 afforested landscape and are legacies of abandoned agriculture (Foster 1992). There is a wide range  
105 of inter- and intra-annual climate variability and extremes among local old field sites across the  
106 region (Oregon State University 2002; Rosenblatt *et al.* 2016, 2019). Vegetation in these old fields  
107 is comprised predominantly of forbs (*Solidago rugosa*, *S. altissima*, *Aster spp*, and grasses (*Poa*  
108 *pratenses*, *Phleum pratense*, *Bromus inermis*, *Agropyron repens*, and *Agrostis spp*; Beckerman  
109 2002; Britton & Brown 1970; Schmitz 2008a). The old fields provide habitat for *Melanoplus*  
110 *femurrrubrum* (hereafter grasshoppers), a moderately sized (2 - 3 cm) generalist herbivore  
111 (Beckerman 2002) that is widely distributed across North America (Helfer 1987). Extensive land  
112 use and development have fragmented the grasshopper's habitat, creating separate populations  
113 throughout its range (Bomar 2001; Parsons & Joern 2014; Rosenblatt *et al.* 2019). The populations  
114 appear to be locally adapted to cope with the climatic conditions of the local ecosystems in which  
115 they reside (Rosenblatt *et al.* 2016, 2019; Baker *et al.* *in review*). This grasshopper species is the  
116 most abundant and persistently present phytophagous insect in the New England old fields  
117 throughout summer and early fall, consuming both grasses and forbs. Its populations also have  
118 discrete generations, in which female grasshoppers deposit eggs in the soil in the late fall before  
119 frost kills all adults, eggs diapause over winter, and juveniles hatch early in the following summer  
120 (Capinera 1987; Chapman & Joern 1991; Uvarov 1977). This combination of discrete, within-

121 population generations makes this grasshopper an ideal study species because it enables clear  
122 attribution of ecosystem effects to population-level trait variation and allows us to examine how  
123 herbivore functional roles interact with biophysical conditions to shape ecosystem functioning.

124 We focused on two functional groups of plants that are important drivers of ecosystem  
125 functioning (Schmitz 2006, 2008a): grasses *Poa spp.* and *Phleum pratense*; and the forb goldenrod  
126 *Solidago rugosa*. These plant functional groups compete asymmetrically, with goldenrod  
127 competitively dominating grasses and other forb plant species in the absence of herbivory (Schmitz  
128 2006). Grasses are nitrogen (N)-rich and are a preferred resource for grasshoppers to build protein  
129 for development, growth and reproduction (Rothley *et al.* 1997). Goldenrod is rich in soluble  
130 carbon (C), which is an essential dietary resource for carbohydrate energy to support metabolism  
131 (Hawlena *et al.* 2012; Rothley *et al.* 1997).

132 This species of grasshopper exhibits both behavioral and physiological plasticity to navigate  
133 local climate variation and environmental stressors. Behaviorally, grasshoppers balance the  
134 benefits and risks of vertical movement within the vegetation canopy: ascending into the upper  
135 canopy provides access to nitrogen-rich forage but also exposes them to higher temperatures and  
136 predation risk, while retreating into the lower canopy reduces these stressors but limits access to  
137 high-quality food (Barton & Schmitz 2009; Pitt 1999). Physiologically, grasshoppers adjust their  
138 metabolism to maintain homeostasis in response to stress. Increased respiration rates under  
139 predation risk heighten the demand for soluble carbohydrate carbon, leading grasshoppers to shift  
140 their diet toward goldenrod, a rich source of soluble C (Schmitz *et al.* 2016; Hawlena & Schmitz  
141 2010).

142 This balance of behavioral and physiological plasticity is not uniform across populations.  
143 Instead, the degree to which each strategy is enlisted varies with the local climate experienced by

144 grasshopper populations across the New England region (Rosenblatt *et al.* 2016, 2019; Baker *et*  
145 *al. in review*). In response to environmental warming, populations occurring in warm sites elevated  
146 their respiration rates and consistently occupied the same vertical stratum of the grassland canopy  
147 throughout the day, potentially because a higher evolved thermal optimum allows them to cope  
148 with warming without behavioral alterations. This combination of heightened metabolic demand  
149 and stable canopy use was accompanied by a shift in diet toward soluble carbon-rich *Solidago*  
150 (Baker *et al. in review*). We term this integrative expression of trait plasticity a climate “resistor”  
151 strategy. In contrast, environmental warming caused populations occurring in cool sites to decrease  
152 their respiration rates and elevate their canopy height, presumably to enhance the likelihood of  
153 reaching their thermal optimum and increasing their access to N-rich forage (Baker *et al. in*  
154 *review*), a strategy we term “reactor”. Together, these trait differences reflect population-level  
155 responses to regional climate variation and underscore how they may shape ecosystem function  
156 under shifting environmental conditions.

157 Grasshopper populations engaging in resistor or reactor strategies stand to have  
158 fundamentally different impacts on the plant community and ecosystem function (Figure S1). A  
159 diet shift by resistors to increase carbohydrate C consumption is expected to reduce goldenrod  
160 abundance, thereby weakening its competitive dominance and allowing other plants, such as  
161 grasses, to increase in biomass (Schmitz 2006). Such herbivory-driven changes in plant species  
162 interactions should cascade to alter the C:N ratios of plant tissue and litter, which can influence  
163 soil microbial respiration and carbon and nitrogen mineralization and ultimately soil C and N  
164 content (Hawlena *et al.* 2012; Hawlena & Schmitz 2010; Schmitz 2006). Alternatively, reactors  
165 that merely feed higher in the canopy without undertaking a diet shift should have weaker  
166 cascading impacts on these properties of the plant community and soil and ecosystem function.

167 *Experimental design*

168 Our three-year field experiment tested the predictions for the effects of resistor and reactor  
169 strategies using grasshopper populations from five old fields sampled across the broader region  
170 (Figure 1). Sites were chosen based on similarity in management, soil properties, hydrology, and  
171 plant community composition, while also capturing a spatial mosaic in mean daily maximum  
172 temperatures. This thermal mosaic was identified in a companion study (Baker *et al. in review*),  
173 which analyzed nine years of remotely sensed climate data (800 m<sup>2</sup> resolution; PRISM Climate  
174 Group 2022) and conducted field-based assessments of grasshopper trait plasticity across eight  
175 populations. Two of the five sites in our study were designated as common gardens for  
176 transplanting populations based on their relative differences in historical climate conditions (warm  
177 vs cool sites) that would lead to resistor and reactor strategies of their local grasshopper  
178 populations.

179 In Year 1, we established replicate, experimental mesocosms to isolate the effect of  
180 herbivores on ecosystem function (Schmitz 2004, 2010; Schmitz *et al.* 2010; Sommer & Schmitz  
181 2020). We used cylindrical mesocosms (0.25 m<sup>2</sup> area × 1 m high) constructed from vinyl-coated  
182 garden wire and wrapped with aluminum insect screening. Each of the five field sites had two  
183 treatments (vegetation-only vs. local herbivores with vegetation) that were replicated eight times  
184 (Figure 1). We further established mesocosms at the two common garden sites to which we  
185 transplanted grasshoppers from the five other populations (Figure 1). Consequently, the entire  
186 experiment included 144 mesocosms. All treatments and populations were randomly assigned to  
187 mesocosm cages.

188 At the beginning of the growing season (June), mesocosm cages were sunk 10 cm into the  
189 soil in each field location, with each mesocosm enclosing natural vegetation. Mesocosm cages

190 were arrayed at least 1 m apart but situated to capture similar initial percent cover in the two  
191 primary functional groups—goldenrod and grass—and to minimize differences in varying  
192 biophysical conditions (e.g., slope, aspect or soil moisture) among mesocosms. Once mesocosms  
193 were established, we removed any resident arthropods and sealed the top with insect screening.  
194 Mesocosms were undisturbed for the growing season to allow ecosystem processes to recover from  
195 the initial disturbance caused by cage installation. At the end of the growing season (October) we  
196 measured baseline ecosystem conditions and functioning (described below) in all mesocosms.  
197 Shortly after the first frost, all mesocosms were “winterized” by removing the insect screen tops  
198 to permit entry of snowfall. This process of clearing the mesocosms of non-target species,  
199 monitoring, and removing the insect screen tops was maintained for all three years of the  
200 experiment.

201 We confirmed similarities in the initial biophysical conditions of the five sites by analyzing  
202 soil samples taken from each field for bulk density, pH, and texture. Bulk density was measured  
203 using soil cores collected from 10 random locations adjacent to the mesocosms. Soil cores were  
204 extracted using a 5 cm × 10 cm AMS Bulk Density Soil Sampler and returned to the lab for  
205 processing. In the lab, each core was sieved to 2 mm. Material < 2 mm was dried at 65°C for 48  
206 hours and weighed. Material > 2 mm was measured for volume displacement. Bulk density was  
207 then calculated as the total dry mass of soil (including < 2 mm and > 2 mm fractions) divided by  
208 the total core volume, accounting for the volume of larger materials measured by displacement.  
209 Soil samples for pH and texture analyses were likewise taken from adjacent locations using the  
210 same soil corer. pH was determined potentiometrically in a slurry system using an electronic pH  
211 meter (Sims & Eckert 2011), while texture was assessed through a standard particle size analysis  
212 (Folk 1966). These soil properties are stable on short timescales and in the absence of human

213 activity (Pahlavan-Rad & Akbarimoghaddam 2018; Xia *et al.* 2020) and were therefore not  
214 measured again in subsequent years.

215 In late June of Year 2, we caught third-instar grasshoppers with sweep nets at their native  
216 sites. We stocked grasshoppers into the mesocosms at the same densities across all field sites,  
217 reflecting the average field density for the species across the region (Rosenblatt *et al.* 2019). For a  
218 given grasshopper population, this meant stocking five individuals in each cage at the homesite  
219 and transplanting another five into each cage at the common garden site(s). One week after  
220 stocking, we replaced any grasshoppers that experienced mortality due to handling stress. After  
221 initial setup, mesocosms were left undisturbed to allow biotic interactions and ecosystem processes  
222 to proceed naturally. During the growing season, we performed non-invasive monitoring by  
223 tapping the mesocosm walls to confirm grasshopper presence and by observing signs of  
224 grasshopper activity, including plant defoliation and frass accumulation. At the end of the growing  
225 season, mesocosms were winterized following the same procedure as Year 1.

226 In late June of Year 3, we again caught third-instar grasshoppers from their native  
227 populations, stocked them into their respective home site and transplant site mesocosms, and  
228 allowed them to progress with external monitoring for the growing season as in Year 2. At the end  
229 of the growing season (October) we again measured ecosystem properties and functioning. We  
230 deliberately did not measure ecosystem properties and functions in Year 2 to avoid altering longer-  
231 term ecosystem functioning that would have arisen from disrupting litterfall and microbial  
232 decomposition, and soil biophysical properties by interim vegetation clipping and soil sampling  
233 (Schmitz 2008b).

234

235 *Ecosystem properties and function measurements*

236            We chose ecosystem properties and function variables based on previous insights about  
237    key old field ecosystem variables that are impacted by grasshoppers (Figure S1; Hawlena et al.  
238    2012; Hawlena & Schmitz 2010; Schmitz 2006, 2008a; Sommer & Schmitz 2020; Sommer et al  
239    *in review*; Strickland et al. 2013) combined with our *a priori* predictions of resistor and reactor  
240    effects explained above. With our goal of disentangling geographic variation, historical conditions,  
241    and the effects of herbivory, we also included initial assessments of soil organic matter in each  
242    mesocosm.

243            At the end of the growing season in Years 1 and 3, we took samples of soil, litter, plant  
244    tissue, and plant percent cover from inside each mesocosm. We took three soil cores (2 cm × 10  
245    cm depth) from inside the mesocosm and placed them together into a plastic bag to homogenize  
246    the samples. All bags were immediately placed on ice and transferred to a refrigerator in the  
247    laboratory on the Yale University campus before being sieved to 4 mm for further processing  
248    (described below). Litter was sampled in three different locations within each mesocosm by taking  
249    material from the surface of the soil. Samples for plant foliar nutrient assessments of goldenrod  
250    and grass were obtained from 10 randomly selected goldenrod leaves and 10 grass clippings  
251    (including stem and inflorescence material, if present) across multiple individuals in each  
252    mesocosm. Each mesocosm's goldenrod and grass plant material was placed into separate coin  
253    envelopes before being dried for 48 h at 60C. Each plant sample was pulverized in a ball mill,  
254    packed in tin capsules and analyzed for C and N using an ECS 4010 Elemental Analyzer (Costech  
255    Analytical Technologies Inc.; Valencia, CA) connected to a Delta Plus Advantage Isotope Ratio  
256    Mass Spectrometer (Thermo Fisher Scientific; Waltham, MA).

257            Plant species percent cover in each mesocosm was estimated by a single experienced  
258    observer to ensure consistency. Plant species were assessed for their spatial extent in a mesocosm

259 and then normalized to 100%, excluding bare ground. We then calculated the Shannon-Wiener  
260 diversity index for each mesocosm using species-specific normalized percentage cover estimates.  
261 We indirectly estimated herbivory effects on the aboveground biomass of plant functional groups  
262 to minimally perturb the mesocosm by creating allometric relationships relating percent cover to  
263 biomass (sensu Schmitz 2003). These allometric relationships were developed from 0.25 m<sup>2</sup> plots  
264 outside the mesocosms, with the proviso that the plots were first selected to target the entire  
265 possible range of percent cover for grasses and goldenrod, before being clipped, dried at 60°C for  
266 48 hours and weighed.

267 We performed several *ex-situ* soil measurements in the lab to capture state and function  
268 variables known to be relevant in this system; specifically, soil %C, soil %N, nitrogen  
269 mineralization, microbial biomass, and soil organic matter. After soils were homogenized and  
270 sieved to 4 mm, they were returned to a refrigerator for subsampling and further processing. All  
271 soil measurements were completed or converted into a shelf-stable form within 2.5 months of field  
272 sampling. Soil subsamples for C and N content analysis were passed through a 2 mm sieve, oven  
273 dried at 60°C for 48 hours, then pulverized in a ball mill, packed into tin capsules and analyzed in  
274 the same manner as the plant and litter material, described above.

275 We calculated the dry-weight equivalent for each mesocosm's soil in terms of gravimetric  
276 moisture content (GWC) and water-holding capacity (WHC). GWC was determined by drying  
277 duplicate 5 g subsamples at 105°C for 24 hours to constant mass and calculating moisture loss.  
278 WHC was measured by saturating duplicate subsamples of sieved soil in a funnel lined with wet  
279 filter paper (Whatman #1), allowing them to drain for 2 hours, and then drying the drained soil at  
280 105°C for 24 hours. The dry weights from these measurements were further used to determine soil  
281 adjustments to 65% WHC for nitrogen mineralization and microbial biomass incubations.

282        Nitrogen mineralization was measured using a 30-day incubation approach with KCl  
283 extraction. Soil was subsampled for Day 1 and Day 30 incubations, with a dry-weight equivalent  
284 of 6 g. For Day 1 extractions, the soil was extracted in 25 mL of 2M KCl. The soil-KCl mixture  
285 was initially mixed by hand, shaken on a shaker-table for 30 minutes, and then refrigerated  
286 overnight to settle. The chilled supernatant was then separated and stored at 4°C until analysis for  
287 colorimetric assays as described below. For Day 30 incubations, the dry-weight equivalent of 6 g  
288 soil was incubated at 20°C in uncovered tubes placed in sealed bags with moist paper towels to  
289 maintain ~65% WHC. Moisture was adjusted weekly based on WHC. On Day 30, the soil was  
290 extracted with 25 mL of 2M KCl following the same procedure as Day 1.

291        Ammonium and nitrate were quantified via colorimetric assays in separate extracts.  
292 Ammonium was analyzed using the salicylate-nitroprusside method (Sims *et al.* 1995), in which  
293  $\text{NH}_4^+$  reacts with hypochlorite and salicylate in an alkaline medium to form a blue indophenol  
294 compound, measured at 660 nm. Nitrate was analyzed via the  $\text{VCl}_3$ /Griess method (Hood-  
295 Nowotny *et al.* 2010), where  $\text{NO}_3^-$  is reduced to  $\text{NO}_2^-$  in a  $\text{VCl}_3$  acidic medium and reacts with  
296 Griess reagents to form a chromophore, measured at 540 nm. N-mineralization rates were  
297 calculated as the changes in ammonium ( $\text{NH}_4^+$ ) and nitrate ( $\text{NO}_3^-$ ) concentrations over time  
298 (difference between the final and initial concentrations in mg/mL), multiplied by the bulk density  
299 (g/cm<sup>3</sup>), and scaling by the soil depth (cm) for units of mg N/cm<sup>2</sup> per month.

300        Soil microbial biomass was estimated using a modified substrate-induced respiration (SIR)  
301 technique (Fierer & Schimel 2003). The equivalent of 4 g dry-weight soil was subsampled from  
302 the soil cores of each mesocosm, incubated overnight at 20 °C, slurried with a 4-mL autolyzed  
303 yeast solution by shaking for 1 h, and then capped with an air-tight lid modified for gas analysis  
304 (Bradford *et al.* 2008). Samples were then flushed with  $\text{CO}_2$ -free air, and after 4 h of incubation at

305 20°C, headspace CO<sub>2</sub> concentrations were measured using an Infra-Red Gas Analyzer (Li-COR  
306 model Li-7000). SIR was then estimated as mg CO<sub>2</sub>-C/g dry soil/h, ensuring standardization across  
307 soil mass and moisture content.

308 Finally, soil organic matter was estimated for each mesocosm using a loss-on-ignition  
309 measurement. Subsamples were passed through a 2 mm sieve and dried at 105 °C before being  
310 placed into a 500 °C furnace for 12 h (Nelson & Sommers 1996). Soil organic matter is a key  
311 determinant of the physical, chemical, and biological processes in soil (Reeves 1997; Robertson  
312 *et al.* 2014; Romig *et al.* 1995), however, unlike the other ecosystem variables, we did not have  
313 any *a priori* hypotheses about the role of herbivory on soil organic matter. Therefore, it was only  
314 measured in the initial year 1 to quantify variability both within and across sites.

315 *Analysis*

316 We employed structural equation modeling (SEM) in R (v.4.4.1) to analyze the direct and  
317 indirect effects of grasshopper herbivory and plasticity on ecosystem properties and function  
318 variables. Due to the nested structure of the data, with measurements collected across five sites,  
319 we implemented a piecewise SEM approach using the piecewiseSEM package (v.2.3.0.1; Lefcheck  
320 *et al.* 2024; Lefcheck 2016). This approach enabled us to incorporate random effects via linear  
321 mixed-effects models (LMMs) using the lme4 package (v.1.1-35.5; Bates *et al.* 2015), ensuring  
322 that site-specific variability was accounted for as a random effect. Specifically, each LMM was  
323 constructed with site as a random effect, categorical treatment (vegetation-only; “resistor”  
324 herbivores; “reactor” herbivores) as a fixed effect, the baseline Year 1 measurement as a fixed  
325 effect, and additional ecosystem variables as fixed effects, where applicable, based on insights  
326 from previous old-field experiments (Hawlena *et al.* 2012; Hawlena & Schmitz 2010; Schmitz  
327 2003, 2006, 2008a; Strickland *et al.* 2013) and assessments of soil ecosystem dynamics (Bradford

328 *et al.* 2008, 2013; Fernández-Martínez *et al.* 2020; Grandy *et al.* 2009; Smith & Bradford 2003).  
329 Rather than model the change in ecosystem variables across years, the baseline Year 1  
330 measurements taken prior to treatment application were included as fixed effects to enable  
331 comparisons between the historic ecosystem state and other predictors. We evaluated model fit for  
332 all LMMs using diagnostics from the DHARMA (v.0.4.6; Hartig 2024) and car (v3.1-3; Weisberg  
333 2019) packages, including assessments of residual normality, homoscedasticity, leverage,  
334 multicollinearity (via variance inflation factors), and model stability. For models that did not meet  
335 all assumptions of residual normality, we used non-parametric bootstrapping (via the boot package,  
336 v.1.3-30; Angelo Canty & Ripley 2024; Davison & Hinkey 1997) as a diagnostic tool to assess the  
337 robustness of fixed effect estimates and 95% confidence intervals. The results of these  
338 bootstrapped models were not used in the SEM itself; instead, they confirmed that the original  
339 models were stable and reliable. We also used the VarCorr function in lme4 to assess the  
340 importance of site as a random effect; where site variability was zero (or near-zero) and model fit  
341 significantly improved, we removed site as a random effect, using a simple linear model instead.  
342 Of the 12 LMMs, 3 met this criterion for simplification.

343 The SEM framework was structured to capture both direct and indirect pathways of  
344 exogenous categorical treatment variables on key response variables. We did not include any latent  
345 variables, as our ecosystem measurements were robust with respect to existing hypotheses. We  
346 evaluated the SEM model fit through global goodness-of-fit statistics, specifically Chi-squared  
347 and Fisher's C, to confirm that our SEM captured the data structure sufficiently. We used directed  
348 separation tests to identify theoretically plausible pathways missing from the initial model. When  
349 the significant pathways identified by these tests aligned with ecosystem theory, they were added  
350 to the respective LMMs, following the same LMM assessment as outlined above. Afterward, the

351 new paths were incorporated into the SEM. In the final model, the statistical significance of  
352 individual predictors was determined based on p-values in the path-specific coefficient summaries.  
353 We also examined marginal and conditional R-squared values from each LMM to assess the  
354 variance explained by fixed predictors alone (marginal R-squared) versus total variance, including  
355 random effects (conditional R-squared). Additionally, individual R-squared values from the SEM  
356 were compared with those from the LMM models, enabling us to disentangle the influence of fixed  
357 effects from site-level variability across treatments. Although a single SEM was used to evaluate  
358 all treatment effects, we present separate visualizations (Figures 2 and 3) to aid interpretability.  
359 These figures reflect subset contrasts extracted from the full model's path estimates, not distinct  
360 SEMs. All data, alongside direct comparisons between two common gardens, can be visualized in  
361 a supplemental Shiny app (<https://nathaliesommer.shinyapps.io/herbivore-populations-structure-ecosystems/>).

363

## 364 **Results**

365 Bulk density, pH, and texture measurements taken during Year 1 site establishment  
366 confirmed comparable soil characteristics across fields. Soil texture ranged between sandy loam  
367 and loam (sand:  $59.94\% \pm 10.17$ ; silt:  $34.02\% \pm 9.63$ ; clay:  $6.10\% \pm 2.39$ ), with an average bulk  
368 density of  $0.69 \text{ g/cm}^3 \pm 0.16$  and a pH of  $5.02 \pm 0.36$ . These results indicate that initial biophysical  
369 conditions were relatively consistent among sites, minimizing potential confounding effects of  
370 baseline soil heterogeneity on experimental outcomes (Figure S2 and S3).

371 Directed separation tests for the SEM supported the inclusion of additional pathways in  
372 seven of the twelve LMMs, predominantly from baseline Year 1 variables. The SEM demonstrated  
373 good overall fit, with a non-significant Chi-squared test ( $\text{Chi}^2 = 187.597$ ,  $\text{df} = 187$ ,  $p = 0.474$ ),

374 indicating that the predicted covariance matrix closely aligned with observed data. Fisher's C was  
375 significant ( $C = 781.807$ ,  $df = 350$ ,  $p < 0.001$ ), as expected in a complex ecological model. Despite  
376 this, individual  $R^2$  values from the SEM closely matched conditional  $R^2$  values from the linear  
377 mixed models (LMMs), confirming the model's robustness in capturing both direct and indirect  
378 effects, as well as site-level variability.

379       Herbivores, in general, played a statistically significant and ecologically meaningful role  
380 in shaping ecosystem responses, particularly by impacting the plant community and soil nutrients  
381 (Figure 2). Across models, herbivore treatments had strong direct effects on goldenrod biomass  
382 (average decrease from vegetation treatments:  $-6.018$ ,  $p < 0.001$ ), grass biomass (average increase  
383 from vegetation treatments:  $+6.851$ ,  $p < 0.001$ ), soil carbon (average decrease from vegetation  
384 treatments:  $-0.171$ ,  $p < 0.001$ ), and soil nitrogen (average decrease from vegetation treatments:  $-$   
385  $0.022$ ,  $p < 0.001$ ). Notably, for these response variables, the effect size of herbivores was greater  
386 than that of historical biophysical conditions or other predictors of ecosystem control such as initial  
387 vegetation biomass, and soil carbon and nitrogen content. Herbivory also had downstream impacts  
388 on foliar nutrients, with grass %C increasing (average increase from vegetation treatments:  $+0.314$ ,  
389  $p < 0.01$ ) and goldenrod %N decreasing (average decrease from vegetation treatments:  $-0.176$ ,  $p$   
390  $< 0.05$ ).

391       The ecosystem impacts of grasshopper populations differed markedly between reactor and  
392 resistor strategies (Figure 3). Reactor herbivores exerted stronger reductions in goldenrod biomass  
393 ( $-7.362$  vs  $-4.675$ ) and slightly smaller reductions in soil carbon ( $-0.157$  vs  $-0.185$ ) than resistor  
394 herbivores. The grasshopper phenotypes had directionally different impacts on nitrogen  
395 mineralization rates, with resistor herbivores causing increased nitrogen mineralization rates  
396 ( $+1.323$ ), and reactor herbivores causing reduced rates ( $-3.066$ ). Indirect ecosystem effects also

397 varied between populations, with resistor grasshoppers decreasing plant diversity (-0.076) and  
398 reactor grasshoppers increasing plant diversity (0.156), following their different impacts on both  
399 goldenrod and grass biomass. These downstream differences reflect the distinct pathways through  
400 which the different herbivore plasticity strategies can influence ecosystem dynamics.

401 Historical condition and site-specific variation did not play predominant roles in ecosystem  
402 dynamics. Baseline Year 1 measurements were significant in six out of twelve response variables  
403 but did not have the largest effect size in any models (Figure S4). Historical conditions, more  
404 generally, were not the predominant variable in the SEM. Historical conditions only had the largest  
405 effect size in two paths for herbivory (grass foliar %C and plant diversity; Figure 2) and two paths  
406 for plasticity (grass foliar %C and goldenrod biomass; Figure 3). For example, grass foliar %C  
407 was most strongly influenced by historical soil %C (effect size: +1.0857,  $p < 0.001$ ), which is  
408 consistent with ecosystem lag effects. Geographic differences were pronounced for some variables  
409 but not for others. Goldenrod biomass (marginal  $R^2 = 0.13$ ; conditional  $R^2 = 0.73$ ; SEM  $R^2 = 0.73$ )  
410 and litter nitrogen (marginal  $R^2 = 0.04$ ; conditional  $R^2 = 0.40$ ; SEM  $R^2 = 0.40$ ) were mainly  
411 influenced by site-level effects, whereas site was removed as a random effect for the soil %C, grass  
412 foliar %C, and %N models due to no explained variability. Other variables reflected substantial  
413 but not predominant geographic heterogeneity, such as soil %N (marginal  $R^2 = 0.30$ ; conditional  
414  $R^2 = 0.54$ , SEM  $R^2 = 0.59$ ), plant diversity (marginal  $R^2 = 0.32$ ; conditional  $R^2 = 0.75$ ), and grass  
415 biomass (marginal  $R^2 = 0.18$ ; conditional  $R^2 = 0.42$ ; SEM  $R^2 = 0.42$ ). Overall, these findings  
416 underscore that while geographic variability and historical ecosystem properties shape ecosystem  
417 function, they do not supersede the effects of animal herbivory or its pathways of effect as  
418 determined at the herbivore population level.

419

420 **Discussion**

421 Our findings demonstrate that differences in local herbivore populations exert a dominant  
422 influence on ecosystem function, surpassing the direct effects of historical conditions and  
423 geographic variation in climate. Across models, herbivores directly altered plant biomass, soil  
424 carbon, and nitrogen availability, restructuring plant communities and modifying nutrient cycling  
425 pathways (Figure 2). While climate-driven geographic variation and historical legacies are widely  
426 recognized as key regulators of ecosystem function (Chapin *et al.* 2011; Reichstein *et al.* 2014;  
427 Schlesinger 2005), our results show that biotic interactions, particularly those mediated by  
428 herbivores, play an equal or even greater role in driving geographic variation in ecosystem  
429 processes (Bardgett & Wardle 2010; Miller *et al.* 2014; Pringle *et al.* 2023). Moreover, differences  
430 among herbivore populations introduced further variability in ecosystem function, leading to  
431 divergent effects on plant composition and nutrient cycling that exceeded the influence of  
432 historical conditions or geography (Figure 3).

433 Herbivore-driven effects on ecosystem function were not only strong but also propagated  
434 through feedback mechanisms that complicate traditional bottom-up models of plant-herbivore  
435 interactions. Geographic variation in herbivory is frequently attributed to differences in plant  
436 nutrient content, defense traits, and local abiotic constraints—factors that influence herbivore  
437 performance and consumption rates from the bottom up (Bradford *et al.* 2014; Lynn *et al.* 2023;  
438 Lynn & Fridley 2019; Marczak *et al.* 2013). This paradigm has helped illuminate how spatial  
439 variation in edaphic conditions can shape plant-herbivore dynamics, but it tends to cast herbivores  
440 as passive responders to their environment rather than active agents in structuring ecosystems. In  
441 contrast, our findings demonstrate that herbivores themselves can drive substantial top-down  
442 effects on biogeochemical cycles and community composition. By altering goldenrod dominance,

443 foliar nutrient content, and rates of nitrogen mineralization, we found herbivores restructured  
444 plant-soil feedbacks in ways not readily explained by background environmental conditions alone  
445 (Figure 2).

446 Divergent ecosystem impacts observed between herbivore populations further challenge  
447 the assumption that herbivore effects are dictated primarily by climate or plant quality. Herbivore  
448 populations from warmer sites increased nitrogen mineralization rates and reduced plant diversity,  
449 while populations from cooler sites suppressed nitrogen mineralization and increased diversity.  
450 These opposing effects occurred despite similar initial conditions across sites, suggesting that trait  
451 differences among herbivore populations—shaped by environmental history but not reducible to  
452 it—play a central role in mediating ecosystem outcomes. Our findings thus question the  
453 conventional assumption that biophysical constraints primarily determine herbivore impacts on  
454 plant communities (Bradford *et al.* 2014; Dostálek *et al.* 2020; Kuglerová *et al.* 2019; Lynn *et al.*  
455 2023; Lynn & Fridley 2019; Marczak *et al.* 2013). Instead, our results align with emerging  
456 evidence that herbivory impacts arise through dynamic feedbacks between consumers and  
457 vegetation, with herbivores acting not only as recipients of environmental filtering but also as  
458 agents of ecosystem restructuring (Barbero-Palacios *et al.* 2024; Maron *et al.* 2014; Massad *et al.*  
459 2024; Pennings & Silliman 2005; Schmitz & Trussell 2016). By shifting competitive hierarchies  
460 among plants and modifying nutrient flows, herbivores contribute to context-dependent patterns  
461 in ecosystem function that are not easily predicted by climate or resource gradients alone. These  
462 findings underscore the importance of integrating top-down consumer-driven processes into  
463 ecosystem models and suggest that herbivore populations may amplify, dampen, or even reverse  
464 the expected effects of environmental change on nutrient cycling and plant community  
465 composition. While our design incorporated spatial replication and extended across three years,

466 the inference space remains limited by the discrete nature of mesocosms and time-point  
467 measurements. Broader-scale, continuous monitoring would be valuable to assess how transient  
468 or persistent these population-level effects are across heterogeneous landscapes.

469 Even within these constraints, the strength of the observed biotic feedbacks raises  
470 important questions about how much such interactions can interact with, or even override, other  
471 widely cited drivers of ecosystem variability, such as historical legacies. While historical legacies  
472 are often considered dominant factors in shaping ecosystem trajectories, (Anderegg *et al.* 2015;  
473 Kulmatiski *et al.* 2006), they did not emerge as dominant predictors of geographic variation in  
474 ecosystem functioning in our study. Baseline Year 1 measurements, made prior to herbivore  
475 manipulation, were statistically significant in six of the twelve models, yet in no case did historical  
476 legacies explain the largest proportion of variation in ecosystem responses. Similarly, geographic  
477 variability influenced some ecosystem variables, particularly goldenrod biomass and litter  
478 nitrogen, but its effects were inconsistent and often weaker than those of herbivory and herbivore  
479 population-level trait differences. Together, this suggests that while historical context may  
480 establish the initial conditions, contemporary biotic interactions can exert stronger influences on  
481 ecosystem function, particularly when consumers differ in their traits like behavior.

482 The distinct functional differences among herbivore populations underscore the  
483 importance of considering population-level variation when evaluating ecosystem processes. In  
484 population ecology, intraspecific variation is often characterized as continuous trait variation  
485 among individuals within a species (Pichon *et al.* 2022; Des Roches *et al.* 2018), and such variation  
486 can be incorporated into models using means and variances in parameters such as consumption  
487 rates or metabolic costs. Conventional consumer-driven ecosystem models typically assume that  
488 individual-level trait variation scales predictably to population-level effects (Evangelista *et al.*

489 2017; Govaert *et al.* 2024; Pichon *et al.* 2022; Raffard *et al.* 2019, 2023). However, our findings  
490 demonstrate that *population-level* trait differentiation may introduce qualitatively distinct  
491 functional effects that are not easily captured by simple continuous distributions. While discrete  
492 population-level categories are one way to represent these shifts, we recognize that structured  
493 continuous functions such as state-dependent trait distributions could also accommodate this  
494 complexity while preserving within-population variance. The divergent ecosystem impacts of the  
495 herbivore populations in our study suggest that models of consumer-driven ecosystem functioning  
496 may need to advance beyond trait means to explicitly account for state-dependent population-level  
497 variation in trait expression (Schmitz & Trussell 2016).

498 As regional climates shift, understanding how trait-based population differences influence  
499 ecosystem processes is critical for predicting ecosystem resilience and anticipating changes in the  
500 functional roles of consumers (Huang *et al.* 2017; Melles *et al.* 2011; Williams & Blois 2018). If  
501 populations exhibit unique trait-based effects on nutrient cycling and plant competition, then  
502 conservation efforts may need to shift from focusing solely on species-level diversity to explicitly  
503 considering population-level functional differences (Allendorf *et al.* 2010; Flanagan *et al.* 2018).  
504 The loss of functionally distinct populations due to habitat fragmentation or climate-driven range  
505 shifts could lead to disproportionate disruptions in ecosystem function. Conservation strategies  
506 might benefit from maintaining a portfolio of functionally diverse populations, ensuring that  
507 ecosystems retain the capacity to mediate nutrient cycling and plant competition under shifting  
508 environmental conditions. Protecting intraspecific diversity may further require targeting  
509 populations that contribute disproportionately to ecosystem resilience rather than assuming all  
510 populations within a species fulfill equivalent functional roles.

511 By resolving the relative contributions of biophysical constraints and consumer-driven  
512 interactions to geographic variation in ecosystem function, this study offers empirical insights into  
513 how to develop a predictive framework for mechanisms creating geographic variation in the  
514 ecosystem consequences of herbivore-plant dynamics, which is a need made increasingly urgent  
515 by ongoing global change. Herbivores, through their population-specific trait differences, emerge  
516 as central regulators of ecosystem function, rivaling or exceeding the influence of historical  
517 legacies and geographic variation in climate. These findings reinforce the need to incorporate  
518 consumer-driven processes into ecological models and conservation planning, particularly as  
519 environmental change reshapes population distributions and functional diversity. Future research  
520 should continue to explore how herbivore-driven processes interact with biophysical constraints  
521 to shape geographic variation in ecosystem function, with an emphasis on how population-level  
522 trait differences mediate ecosystem resilience in changing environments.

523 **Figure captions**

524 **Figure 1.** Our field experiment was conducted over three growing seasons to assess the effects of  
525 population-level variation in grasshoppers on ecosystem function across multiple sites in New  
526 England, USA. In Year 1, field mesocosms were established, and baseline ecosystem conditions  
527 were measured at each site (black circles with field labels corresponding to site-level data  
528 presented in Figures S1 and S2). In Year 2, grasshoppers were reciprocally transplanted between  
529 warm-origin (gold, “resistor” populations) and cool-origin populations (purple, “reactor”  
530 populations) with individuals stocked at their homesites (circles) and common gardens (stars).  
531 Throughout the second growing season, mesocosms were monitored and maintained to ensure  
532 population establishment. In Year 3, the transplant was repeated and final ecosystem  
533 measurements were taken to quantify the effects of different herbivore populations on plant  
534 biomass, soil nutrients, and nitrogen mineralization. The study design allowed for direct  
535 comparison of how functionally distinct grasshopper populations mediate ecosystem function  
536 under variable environmental conditions.

537

538 **Figure 2.** The path diagram illustrates the statistically significant ( $p < 0.05$ ) direct and indirect  
539 effects of vegetation-only treatments and the average herbivory treatment on key ecosystem  
540 response variables. Arrow widths are scaled to represent effect sizes, with positive effects shown  
541 as solid lines and negative effects as dashed lines. The values next to each arrow indicate the effect  
542 size. To simplify the visualization, historical ecosystem variables (i.e., the Year 1 baseline metrics)  
543 are shown in gray, but were only included in the path diagram when their effect size was the largest  
544 for that respective model. A full path diagram with all significant paths regardless of the effect size  
545 can be found in Fig. S3. SEM results were visualized using the DiagrammeR package (v.1.0.11).

546

547 **Figure 3.** The path diagram illustrates the statistically significant ( $p < 0.05$ ) direct and indirect  
548 effects of different forms of herbivore plasticity on key ecosystem response variables. All paths  
549 are derived from a single SEM model; however, the values shown here reflect the difference in  
550 effect size between the two herbivore treatments relative to vegetation-only controls (i.e., the net  
551 difference in effect size between the herbivore treatment and the vegetation treatment). As in  
552 Figure 2, arrow widths are scaled to represent effect sizes, with positive effects shown as solid  
553 lines, negative effects show as dashed lines, and the value next to each arrow indicates the effect  
554 size. Historical ecosystem variables (i.e., the Year 1 baseline metrics) were only included in the  
555 path diagram when their effect size was the largest for that respective model, shown in gray nodes  
556 and black arrows.

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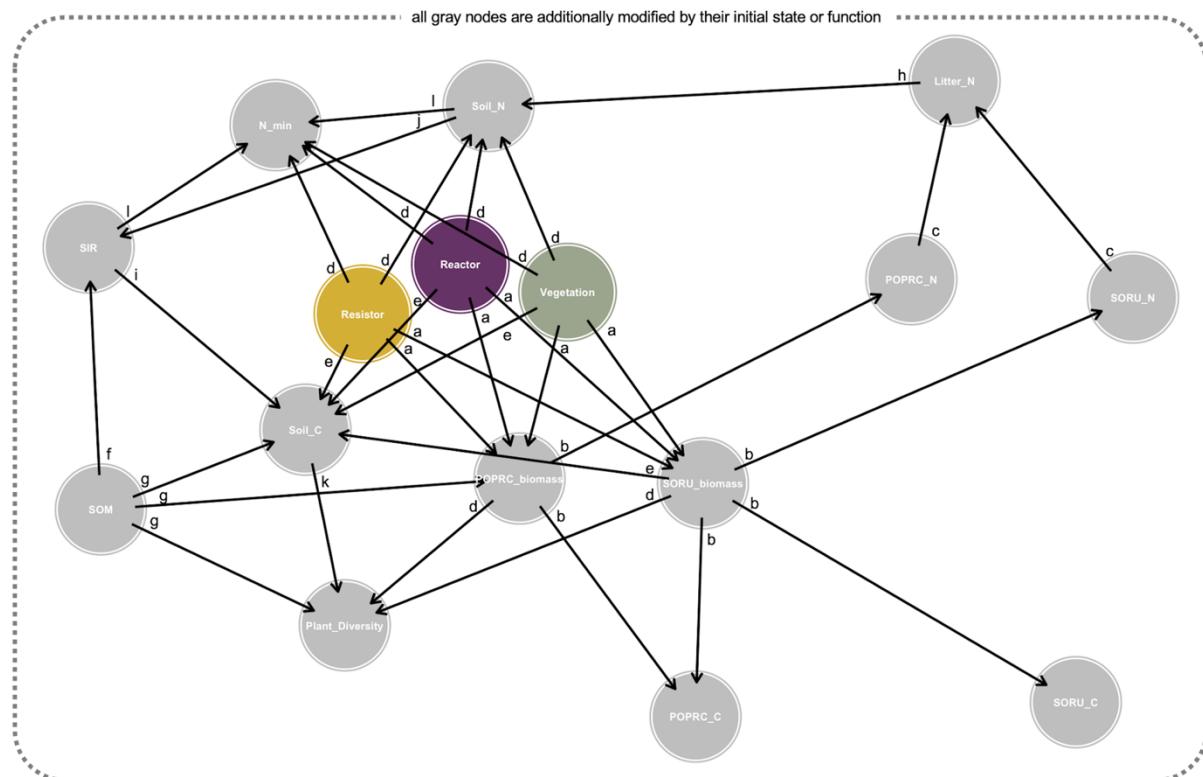
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 809 **Supporting Information for Sommer *et al.* Herbivore population differences rival**  
 810 **geographic and biophysical variation in structuring ecosystem function**

812 **Figure S1.** Directed acyclic graph of hypothesized pathways for herbivore and vegetation  
 813 effects tested in the structural equation model. All paths were grounded in prior empirical  
 814 work, signified by the letter from the originating node.



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 816 <sup>a</sup>Baker et al. in review; Rosenblatt et al. 2019; Schmitz 2003, 2008; <sup>b</sup>Hawlena & Schmitz 2010;  
 817 Rothley et al. 1997; Schmitz et al. 2016; <sup>c</sup>Hawlena et al. 2012; Schmitz 2006; <sup>d</sup>Bradford et al. 2013;  
 818 Schmitz 2006; <sup>e</sup>Fernández-Martínez et al. 2020; Schmitz 2006; Schmitz et al. 2017; Strickland et al.

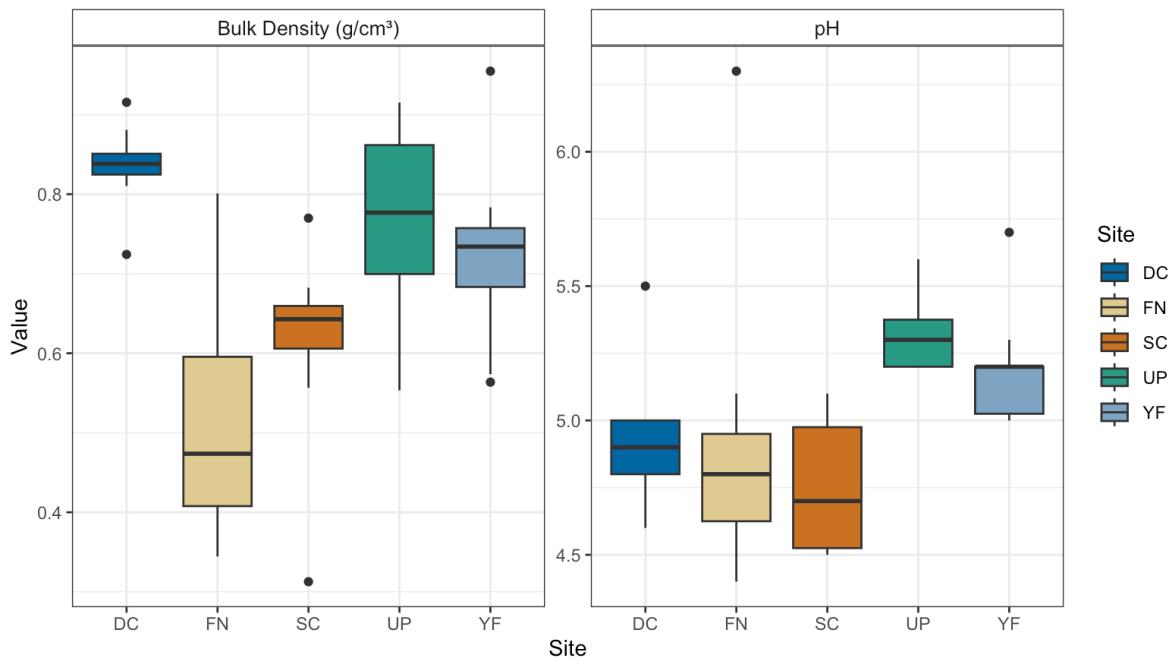
819 2013; <sup>f</sup>Bradford et al. 2013; Grandy et al. 2009; <sup>g</sup>Grandy et al. 2009; Reeves 1997; <sup>h</sup>Smith &  
820 Bradford 2003; <sup>i</sup>Tao et al. 2023; <sup>j</sup>Dong et al. 2022; <sup>k</sup>Anacker et al. 2021; Chen et al. 2018; <sup>l</sup>Liu et al.  
821 2016; Xu et al. 2024

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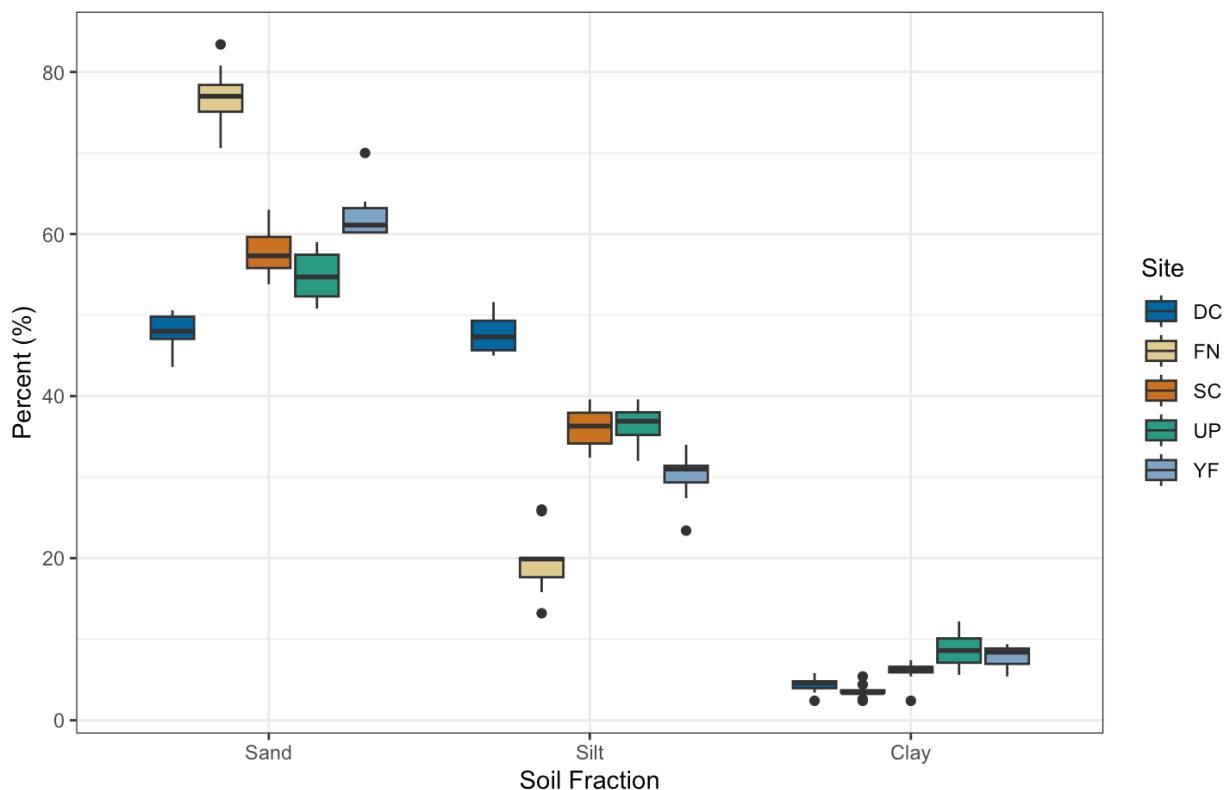
**Figure S2.** Bulk density and pH across all five sites, measured in Year 1. Site abbreviations correspond to the map in Figure 1 (main text).



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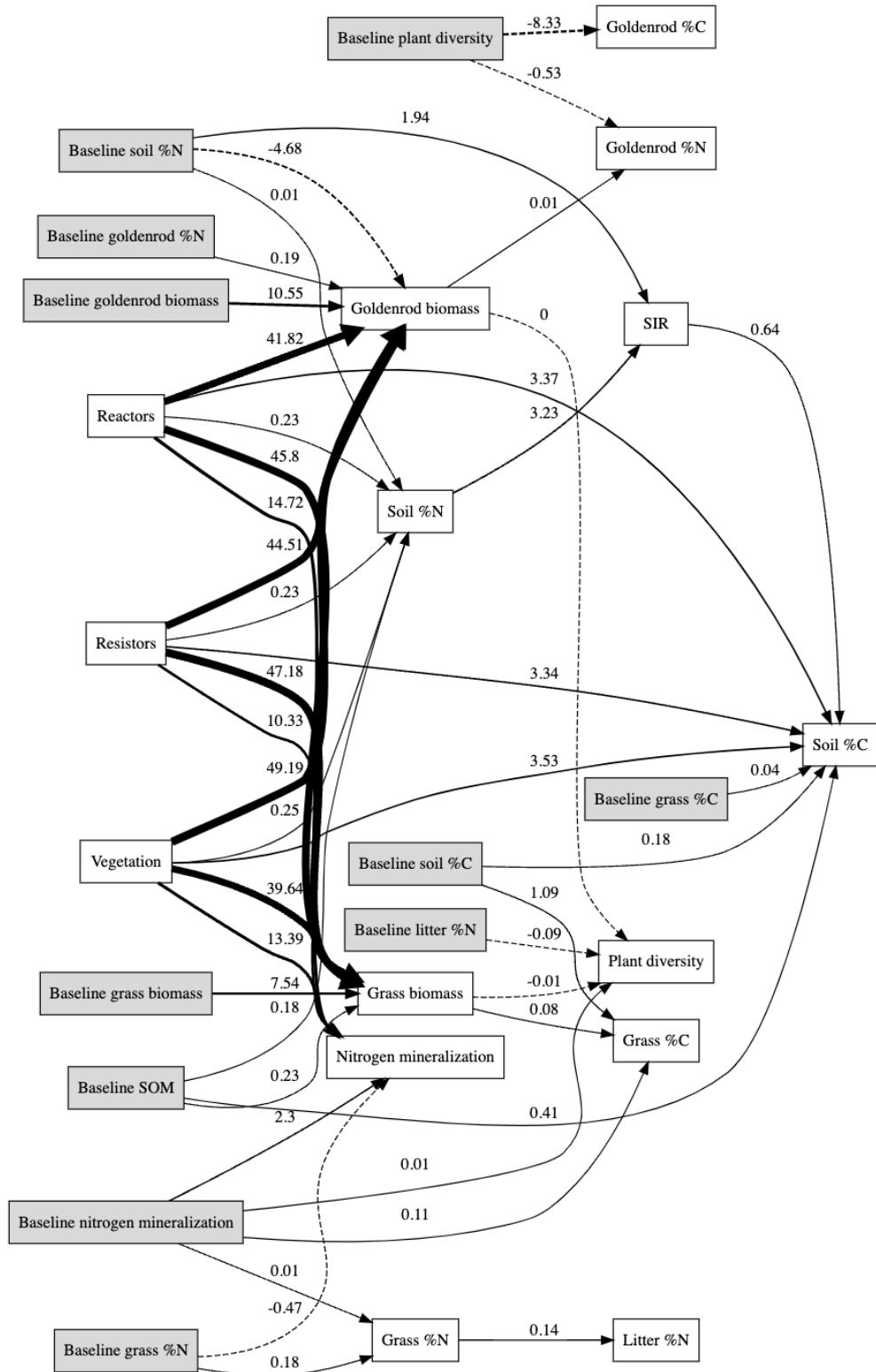
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**Figure S3.** Soil texture components measured across all five sites in Year 1. Site abbreviations correspond to the map in Figure 1 (main text).



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**Figure S4.** Full SEM; includes all statistically significant paths [ $p < 0.05$ ])



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