

Running head: Burn severity and ecosystem transformation

Title: Fuel connectivity, burn severity, and seedbank survivorship drive ecosystem transformation in a semi-arid shrubland.

Adam L. Mahood^{1,2,3,*}, Michael J. Koontz², Jennifer K. Balch^{1,2,4}

¹ Department of Geography, University of Colorado Boulder, Boulder, CO, USA

² Earth Lab, University of Colorado, Boulder, CO, USA

³ Water Resources, Agricultural Research Service, United States Department of Agriculture, Fort Collins, CO, USA

⁴ Environmental Data Science Innovation and Inclusion Lab, University of Colorado, Boulder, CO

* Corresponding author: admahood@gmail.com

Open Research Statement: Data and code to recreate the analysis are freely available at <https://doi.org/10.5281/zenodo.5293996>.

1 **Abstract**

2 A key challenge in ecology is understanding how multiple drivers interact to precipitate
3 persistent vegetation state changes. These state changes may be both precipitated and
4 maintained by disturbances, but predicting whether the state change is fleeting or persistent
5 requires an understanding of the mechanisms by which disturbance affects the alternative
6 communities. In the sagebrush shrublands of the western United States, widespread annual
7 grass invasion has increased fuel connectivity, which increases the size and spatial contiguity
8 of fires, leading to post-fire monocultures of introduced annual grasses (IAG). The novel
9 grassland state can be persistent, and more likely to promote large fires than the shrubland
10 it replaced. But the mechanisms by which pre-fire invasion and fire occurrence are linked to
11 higher post-fire flammability are not fully understood. A natural experiment to explore these
12 interactions presented itself when we arrived in northern Nevada immediately after a 50,000
13 ha wildfire was extinguished.

14 We hypothesized that the novel grassland state is maintained via a reinforcing feedback
15 where higher fuel connectivity increases burn severity, which subsequently increases post-fire
16 IAG dispersal, seed survivorship, and fuel connectivity. We used a Bayesian joint species
17 distribution model and structural equation model framework to assess the strength of the
18 support for each element in this feedback pathway. We found that pre-fire fuel connectivity
19 increased burn severity and that higher burn severity had mostly positive effects on the
20 occurrence of IAG and another non-native species, and mostly negative or neutral relationships
21 with all other species. Finally, we found that the abundance of IAG seeds in the seedbank
22 immediately post-fire had a positive effect on the fuel connectivity 3 years after fire, completing
23 a positive feedback promoting IAG. These results demonstrate that the strength of the positive
24 feedback is controlled by measurable characteristics of ecosystem structure, composition
25 and disturbance. Further, each node in the loop is affected independently by multiple
26 global change drivers. It is possible that these characteristics can be modeled to predict

27 threshold behavior and inform management actions to mitigate or slow the establishment of
28 the grass-fire cycle, perhaps via targeted restoration applications or pre-fire fuel treatments.
29 *Keywords:* *Artemisia tridentata*, alternative stable states, *Bromus tectorum*, burn severity,
30 cheatgrass, fuel connectivity, grass-fire cycle, joint species distribution model, resilience,
31 sagebrush

32 **1. Introduction**

33 Ecosystems around the world are being affected simultaneously by multiple facets of global
34 change. For example, changes in land use can facilitate exotic plant invasions (Allan et al.
35 2015), which can alter ecosystem structure (Davies and Nafus 2013). Altered structure can
36 change the likelihood of a disturbance, the properties of a disturbance and the capacity
37 of the system to recover after a disturbance (Brooks et al. 2004). Global climate change
38 can also directly affect the magnitude of disturbances (S. A. Parks and Abatzoglou 2020),
39 and act as a demographic filter that influences how ecosystems recover after disturbances
40 (Rother, Veblen, and Furman 2015; Davis et al. 2019) via impacts on adult plant survival
41 and seed dispersal (Davis, Higuera, and Sala 2018; Eskelinen et al. 2020). The combined
42 effects of global change forces on structure, function and disturbance can cascade and interact.
43 For example, while burn severity (or the proportion of biomass burned (Keeley 2009)) is
44 influenced by vegetation structure (Koontz et al. 2020; Sean A. Parks et al. 2018), it also
45 increases with temperature and aridity (S. A. Parks and Abatzoglou 2020). These forces
46 can ultimately lead to permanent compositional change, biodiversity losses and the loss of
47 ecosystem services (Ratajczak et al. 2018; Mahood and Balch 2019; Mahood et al. 2022)
48 due to internal, self-reinforcing mechanisms that arise from those structural and functional
49 changes which then maintain an alternative stable state (Marten Scheffer and Carpenter
50 2003; Ratajczak et al. 2018).

51 There is a long history of univariate time series observations that show sudden state changes

52 (Marten Scheffer and Carpenter 2003), and these have informed the development of the-
53 ories that help us understand how systems of any type can change state suddenly, and
54 exist in persistent alternative stable states (Marten Scheffer et al. 2015; Ratajczak et al.
55 2018). These theories typically represent the system’s state with a single variable, of which
56 the mean is observed to abruptly change in time or space (Marten Scheffer et al. 2015).
57 Descriptive evidence of alternative stable states has been documented at broad scales in
58 tropical ecosystems, where forests, savannas and grasslands are considered alternative stable
59 states because they are floristically distinct (Aleman et al. 2020) and cluster around static
60 values of woody cover (80, 30 and 0 percent) while occurring along overlapping ranges of
61 precipitation (Hirota et al. 2011; Staver, Archibald, and Levin 2011). The forested state has
62 a self-reinforcing, positive feedback between evapotranspiration and tree cover (Staal et al.
63 2020), while the grassland and savanna states are maintained by feedbacks between grass
64 flammability and fire occurrence (D’Antonio and Vitousek 1992; Staver, Archibald, and Levin
65 2011). Alternative stable states are believed to be widespread (M. Scheffer et al. 2001), but
66 their existence is rarely proven at broader scales, with most demonstrative studies having
67 been conducted in greenhouse and laboratory microcosm experiments (Schröder, Persson,
68 and De Roos 2005). One of the reasons for this is that ecological systems are much more
69 complex than a simple bivariate system with a single driver and a single response. There
70 may be multiple drivers, and the state is the product of interactions between organisms and
71 their immediate environment, as well as countless inter- and intra-specific interactions.

72 A central challenge in ecology in the 21st century is to move from describing how plant
73 communities are affected by global change to the capacity to predict how species pools will
74 assemble and persist in response to global change (Davis, Higuera, and Sala 2018; Keddy
75 and Laughlin 2021). Prediction of community response to multi-faceted global change
76 drivers is enhanced with a better understanding of the mechanisms that underlie community
77 stability in the face of disturbances. A classic example of an ecosystem that appears to have
78 disturbance-mediated alternative stable states (but see Morris and Leger (2016)), but whose

79 stability mechanisms aren't well understood is the invasion of *Bromus tectorum* L. and other
80 introduced annual grasses in the Great Basin of the western United States. Here, it is well
81 documented how the interaction of annual grass invasion, fire (Balch et al. 2013) and grazing
82 (Williamson et al. 2019) are associated with the degradation or loss of over half of Wyoming
83 big sagebrush (*Artemisia tridentata* ssp. *wyomingensis* Beetle & Young) ecosystems (Davies
84 et al. 2011). These systems had a precolonial fire regime of infrequent, patchy fires (Bukowski
85 and Baker 2013). In uninvaded areas, the space between shrubs is typically composed of bare
86 ground covered in biological soil crust and caespitose perennial plants (Figure 1). Because
87 fire does not spread readily below a threshold of approximately 60% cover of flammable
88 vegetation (Archibald, Staver, and Levin 2012), the low fuel connectivity in these areas limits
89 fire spread. Annual grass invasion increases fuel connectivity while decreasing fuel moisture
90 (Brooks et al. 2004; Davies and Nafus 2013), leading to increased fire size and frequency
91 (Balch et al. 2013). Sagebrush stands with high native perennial cover might need only a
92 small amount of additional annual grass cover to alter ecosystem structure enough to alter the
93 fire regime (Figure 2). After fire, the landscape is typically dominated by introduced annual
94 grasses. But in order to understand how fire drives the persistence of the grassland state, we
95 need to understand the demographic mechanisms by which fire impacts propagule dispersal
96 and benefits the alternative state (Davis, Higuera, and Sala 2018). As with forested systems,
97 propagule dispersal is a key filter through which species must pass in order to establish and
98 persist in a post-fire landscape (Gill et al. 2022).

99 Petraitis and Latham (1999) posited that the maintenance of alternate species assemblages
100 requires first a disturbance that removes the species from the initial assemblage and second
101 the arrival of the species of the alternate assemblage. One understudied mechanism that
102 may explain both for the *Artemisia/Bromus* system is the interaction between the species
103 composition of the soil seed bank and burn severity. Because the invading species are annual,
104 and many of the key native plant species are seed obligates, the seed is the key life history stage
105 that fire must act upon to benefit the invading plants. Seeds and seedlings are particularly

106 vulnerable to climate, competition and disturbance (Enright et al. 2015). Warmer and drier
107 conditions simultaneously reduce recruitment, growth, and survival of seeds and seedlings
108 (Enright et al. 2015; Schlaepfer, Lauenroth, and Bradford 2014), while also increasing burn
109 severity (S. A. Parks and Abatzoglou 2020). In fire prone ecosystems, seed obligate species
110 typically have life history strategies to cope with fires that burn at different severities (Maia
111 et al. 2012; Wright, Latz, and Zuur 2016; Palmer, Denham, and Ooi 2018). Soil heating from
112 fire affects the response of vegetation to fire (Gagnon et al. 2015), including the capacity of
113 seeds to remain viable after fire (Humphrey and Schupp 2001). High severity fire can affect
114 species that use the seedbank positively (Kimura and Tsuyuzaki 2011), negatively (Heydari
115 et al. 2017), or have no effect (Lipoma, Funes, and Díaz 2018), depending on species-specific
116 adaptations. Both the depth of the burn and fire temperature can affect subsequent recovery
117 by seed germination (Morgan and Neuenschwander 1988; Schimmel and Granström 1996), as
118 well as seed mortality and physical seed dormancy mechanisms (Liyanage and Ooi 2017).

119 In addition to size and frequency, exotic plant invasions can alter fire temperature (Brooks et
120 al. 2004; R. O. Jones et al. 2015) and burn severity. While in many cases fires that burn
121 at higher temperatures will also consume more biomass (i.e. burn at higher severity), grass
122 fires may not always have such a relationship. Direct measurements have shown that *B.*
123 *tectorum* burns at low temperatures (Beckstead et al. 2011; Germino, Chambers, and Brown
124 2016), but because it also increases horizontal fuel connectivity (Davies and Nafus 2013), it
125 leads to more contiguously burned areas and therefore higher burn severity, despite lower fire
126 temperatures. To benefit from fire, *B. tectorum* would need to gain a fitness benefit relative
127 to other species

128 One way to achieve this is to disperse more viable seeds into the post-fire landscape than
129 the other species and become well-represented in the post-fire plant assemblage (Bond and
130 Midgley 1995). If the fire is patchy, this can happen through post-fire seed dispersal (Monty,
131 Brown, and Johnston 2013). Without unburned patches, seeds must survive the fire. If
132 the increase in fuel connectivity caused by *B. tectorum* increases the severity of fire, one

133 way burn severity might then influence the community composition of the post-fire seed
134 bank to facilitate the post-fire dominance of *B. tectorum* would be to burn a contiguous
135 area at a temperature high enough to kill fire-intolerant native seeds, but low enough that
136 *B. tectorum* seeds survive and germinate more readily from fire-induced germination cues
137 (Naghipour et al. 2016; Fenesi et al. 2016). In other words, an area with high burn severity
138 should have a lower relative occurrence of viable seeds of native species, and a higher relative
139 occurrence of the seeds of fire-tolerant introduced annual plants. This would allow for the for
140 the often-observed dominance of introduced annual grasses after a few years and would result
141 in higher fuel connectivity, closing the positive feedback loop. Plants that are not adapted
142 to frequent fire would be less likely to produce seeds that are adapted to surviving fire, or
143 dispersal mechanisms to take advantage of the resources available immediately after fire
144 (Keeley et al. 2011). To our knowledge, despite several studies on the relationship between
145 fire occurrence and the seed bank in this system (Hassan and West 1986; Humphrey and
146 Schupp 2001; Boudell, Link, and Johansen 2002), no studies to date have examined the effect
147 of burn severity on the seed bank. Burn severity is more ecologically meaningful than fire
148 occurrence, and is more useful for understanding threshold effects and stable states than a
149 binary variable.

150 Here, we collected soil cores from 14 locations along the perimeter of a large fire (the Hot
151 Pot fire, ~50,000 ha) immediately after it was extinguished, in northern Nevada in July
152 2016. Each location had paired burned and unburned samples. Because it burned a large
153 area in only three days, we could sample a broad area while being reasonably certain that
154 the weather conditions during the fire were similar at all sites. Because we collected our
155 samples immediately after the fire was extinguished, we felt confident that the seed bank
156 samples did not contain seeds deposited by post-fire dispersal. We put the samples in cold
157 storage and germinated the seeds from those cores in a greenhouse the following spring. In
158 spring 2017 and fall 2019 we collected information on vegetation structure and diversity at
159 each location. We tested three hypotheses in this study that are depicted in Figure 3: (H1)

160 Pre-fire fuel connectivity would be positively related to burn severity; (H2) burn severity
161 would increase the occurrence probability of introduced annual species in the seed bank and
162 reduce the occurrence probability of native species; and (H3) the abundance of post-fire
163 *B. tectorum* seeds in the seedbank would be positively related to post-fire fuel connectivity.
164 We examined two alternatives to H2: (H2a), increased fuel connectivity brought on by the
165 invasion of annual grasses may have already depleted the diversity of the soil seed bank
166 before the fire occurred; and (H2b) prefire fuel connectivity is solely reflective of annual grass
167 cover, which drives post-fire annual grass seed abundance. In addition, because in our study
168 system post-fire sites are floristically distinct from the pre-fire state (Mahood and Balch
169 2019), typically with near monocultures of *B. tectorum*, we hypothesized that (H4) high
170 post-fire fuel connectivity of those near-monocultures would result in lower aboveground
171 species diversity due to competitive exclusion of native plants.

172 2. Methods

173 2.1 Study Area

174 The study was conducted in north-central Nevada the day after a large fire (the Hot Pot Fire)
175 was extinguished (Appendix S1, Fig. S1). The Hot Pot Fire burned just over 50,000 hectares
176 in less than a week. The pre-fire landcover was predominantly *B. tectorum* and Wyoming big
177 sagebrush plant communities. The fire occurred after the early season plants, including *B.*
178 *tectorum* and *Poa secunda* J. Presl, the most abundant native understory species, had gone
179 to seed, and before the late season species, including Wyoming big sagebrush, had produced
180 flowers. Thus we were able to isolate the effect of the fire without any confounding effects of
181 post-fire seed dispersal, while achieving a broad spatial extent. The sites we sampled ranged
182 from 1,397 to 1,607 meters in elevation.

183 2.2 Seed Bank Sampling

184 In early July 2016, we collected samples of the soil seed bank at fourteen locations the day

185 after the Hot Pot fire was contained. Each site was located at the perimeter of the fire where
186 it was clearly delineated by a bulldozer line or in one case a narrow dirt road. We were
187 confident paired sites were of the same pre-fire composition because we had been working in
188 these areas all summer collecting data for another study. Eleven sites were mature sagebrush
189 communities with no history of fire since at least 1984. Three sites had previously burned in
190 1984 according to the Monitoring Trends in Burn Severity (MTBS) fire history ([Eidenshink
191 et al. 2007](#)) and had high cover of *B. tectorum*, but still had scattered sagebrush cover. We
192 used a metal stake to mark paired burned and unburned sampling locations on each side of
193 the perimeter, 10 m from the nearest evidence of anthropogenic disturbance (i.e. bulldozer
194 effects, footprints) associated with active fire suppression along the perimeter. Within 3 m of
195 each marker, we extracted twelve, 6 cm deep, 5 cm diameter, soil cores. Seeds of sagebrush
196 generally do not fall far (<30 m) from their parent plants in this system ([Shinneman and
197 McIlroy 2016](#)), and so they are not uniformly distributed ([Boudell, Link, and Johansen 2002](#)).
198 In addition, seeds from *B. tectorum* and *Artemisia* have different germination rates based
199 on the micro-site they find themselves in (i.e. under a shrub or in the bare ground between
200 shrubs, [Eckert et al. 1986](#)). To account for these potentially confounding effects, we placed
201 half of the core locations under shrubs, half in shrub interspaces, and aggregated the cores
202 for each site. In the burned areas, it was obvious where shrubs had been located. Even
203 when they were completely incinerated, their imprint remained on the soil surface ([Bechtold
204 and Inouye 2007](#)). To examine the effect of seed depth, we divided each soil core into 0-2
205 cm and 2-6 cm depths. Litter was aggregated with the 0-2 cm samples. Samples were then
206 placed in cold storage (~2 deg C) for 3 months ([Meyer, Monsen, and Mearthur 2013](#)). At all
207 sites, to be sure that we were at a site where sagebrush germination could occur we checked
208 for first year germinants on the unburned side (we found them at all sites), and to ensure
209 that there were no confounding effects of post-fire seed dispersal, we determined whether or
210 not the sagebrush were flowering (they were not flowering at all sites), and recorded species
211 occupancy for all aboveground plant species.

212 We followed the methodology of Ter Heert et al. (1996) to germinate the seeds. Each sample
213 was run through 0.2 mm sieve, and spread in a 3-5 mm layer over the top of 1 - 4 pots.
214 These pots were filled 3 cm deep with potting soil, topped by a thin layer of sand. Pots were
215 watered as needed to stay at field capacity. Every week emerging germinants were identified,
216 counted and removed. Most of the germination occurred within 6 weeks, and after 8 weeks
217 we ended the germination assay.

218 *2.3 Post-Fire Vegetation Sampling*

219 We sampled the aboveground fuel structure and plant diversity in May 2017, the growing
220 season immediately after the fire and again in September 2019. At each location, we established
221 50m transects starting at the boundary of the burned and unburned sides of the perimeter,
222 running perpendicular to the fire perimeter, and marked the transect ends with rebar. In
223 order to characterize aboveground plant diversity, we measured the occupancy and abundance
224 of all plant species by measuring cover of every species in 0.1 m² quadrats spaced every 5
225 m along each transect. We measured shrub cover (coarse fuels) and herbaceous plant cover
226 (fine fuels) using the line intercept method along the transect, a commonly-used approach for
227 characterizing fuel structure (Elzinga, Salzer, and Willoughby 1998). We calculated total
228 vegetation cover (TVC) as the sum of the fine and coarse fuel measurements. Both live and
229 dead plants were included in these measurements.

230 *2.4 Remotely-Sensed Burn Severity*

231 We downloaded the “fire bundle” of the Hot Pot fire from www.mtbs.gov. This included
232 cloud-free Landsat 8 scenes collected before the Hot Pot fire, and already calculated layers of
233 the Differenced Normalized Burn Ratio (dNBR, Equations 1 & 2, J. D. Miller et al. 2009).
234 Because our sites were generally within 10 meters of the burn perimeter, The pixels directly
235 intersecting the site locations were likely to be mixed pixels (i.e. containing burned and
236 unburned ground). To minimize this effect, we extracted all the dNBR values within a 120
237 meter buffer of each seed bank site for pixels whose centroids fell inside of the fire perimeter

238 and calculated the mean.

239 **Equation 1:** $NBR = (NIR - SWIR_1)/(NIR + SWIR_1)$

240 **Equation 2:** $dNBR = (NBR_{pre\ fire} - NBR_{post\ fire}) * 1000$

241 2.5 Statistical Analysis

242 Our statistical analysis centered around trying to understand each component of the positive
243 feedback loop posited by the 4 hypotheses described above. In order to understand how
244 pre-fire fuel connectivity influenced burn severity (H1), we used total vegetation cover (TVC)
245 from two separate data sources as a proxy for fuel connectivity, and created separate linear
246 models with TVC as the predictor variable and burn severity (dNBR, [J. D. Miller et al.](#)
247 [2009](#)) as the response variable. With the field data we collected, we created an ordinary
248 least squares (OLS) linear model with burn severity as the dependent variable and TVC
249 (defined as shrub cover plus herbaceous plant cover from the unburned side of the paired
250 sites), elevation and aspect as independent variables.

251 We were concerned that because our data were collected at the edge of the fire, the burn
252 severity calculated at each point may have included partially burned pixels. So, as a
253 supplement, we examined the same relationship by creating a model of TVC using Landsat
254 Thematic Mapper (TM) surface reflectance data using field measurements of TVC from the
255 Bureau of Land Management’s Assessment, Inventory and Monitoring dataset (AIM, [U.S.](#)
256 [Department of Interior 2018](#)). The AIM dataset contained 813 sampling locations within
257 the Central Basin and Range ecoregion ([Commission for Environmental Cooperation 2006](#))
258 that were visited by BLM field crews between 2011 and 2015. They were mostly sampled
259 once but there were some repeats, for 1,117 total measurements. For each of these points,
260 we extracted the surface reflectance values of each Landsat band for the sampling year
261 near peak biomass using a cloud-free scene from May or early June. Then, we used those
262 surface reflectance values to calculate various vegetation indexes (Appendix S1: Table S1),
263 including the Green Normalized Difference Vegetation Index (Green NDVI, Equation 3), and

264 Normalized Difference Senesced Vegetation Index (NDSVI, Equation 4). We used these two
265 indexes and their interactions as predictors in a generalized linear model of TVC with a
266 beta distribution. We used the model to create a layer of estimated pre-fire TVC for the
267 study area, and extracted both our predictions of TVC and dNBR of the fire from 1000
268 regularly-spaced points within the fire perimeter. Finally, to quantify the effect of TVC on
269 burn severity, we created an OLS linear model with our modeled TVC and its second-order
270 polynomial as predictor variables and burn severity as the response variable.

271 **Equation 3:** $Green\ NDVI = \frac{NIR-Green}{NIR+Green}$

272 **Equation 4:** $NDSVI = \frac{SWIR_1-Red}{SWIR_1+Red}$

273 To examine how burn severity affected the community composition of the seed bank (H2),
274 we created a joint species distribution model (JSDM) in a Bayesian framework ([Tikhonov et](#)
275 [al. 2020](#)) for the occurrence of all species germinated from the seed bank that were found
276 at more than one location. We created four Markov Chain Monte Carlo (MCMC) chains,
277 each consisting of 150,000 iterations. We discarded the first 50,000 iterations for each chain
278 and then recorded every 100th for a total of 1,000 posterior samples per chain, and 4,000
279 total. We assessed model convergence using the effective sample size and the potential
280 scale reduction factor ([Gelman, Rubin, et al. 1992](#)). We used the model to predict the
281 probability of occurrence of germinable seeds of a given species along a gradient of burn
282 severity. We included burn severity, elevation, aspect, pre-fire seedbank diversity and soil
283 depth as independent variables.

284 To account for the possibility that increased fuel connectivity brought on by the invasion
285 of annual grasses may have already depleted the diversity of the soil seed bank before the
286 fire occurred (H2a) as a confounding factor, we included the Shannon-Weaver diversity
287 index ([Shannon and Weaver 1949](#)) in the paired, unburned seed bank samples as one of the
288 predictor variables in our JSDM. We also created OLS models with the unburned species
289 richness and Shannon-Weaver diversity index predicted by prefire fuel connectivity, with the

290 expectation that pre-fire fuel connectivity would have had a negative effect on the prefire
291 seedbank diversity. To examine how community composition and burn severity then affected
292 subsequent fuel connectivity (H3), we created OLS models with fuel connectivity three years
293 post-fire as the dependent variable, and burn severity, seed counts for *B. tectorum*, *P. secunda*
294 and other species, elevation, aspect, depth, and alpha diversity as independent variables.
295 To examine how the resulting fuel connectivity was related to biodiversity (H4), we used
296 the aboveground diversity data and connectivity data that we collected in 2019 to create a
297 Poisson GLM with number of species encountered at each site as the dependent variable,
298 as well as an OLS linear model with the Shannon-Weaver index for the plant species as a
299 dependent variable. We used fuel connectivity, elevation, and aspect as independent variables.
300 In order to examine hypotheses 1-3 in a single framework we constructed a path model
301 (Rosseel 2012). We had paths leading from pre-fire connectivity, through burn severity to
302 the log of the post-fire count of *B. tectorum* seeds in the seedbank, and finally to post-fire
303 connectivity. Pre-fire cover of *B. tectorum*, elevation, pre-fire seed bank diversity and pre-fire
304 aboveground diversity were also accounted for.

305 All analyses were done in R (R Core Team 2020). Data and code to recreate the analysis are
306 freely available at <https://doi.org/10.5281/zenodo.5293996>.

307 **3. Results**

308 We found support for each hypothesized component of the positive feedback loop independently
309 and when combined in the path model ($\chi^2 = 3.17$, $p = 0.39$, Figure 4a, Appendix S1, Tables
310 S4 & S5). For H1, TVC had a weak positive relationship with burn severity ($\beta = 2.4$, p
311 $= 0.083$, $R^2 = 0.27$, Figure 4b, Appendix S1: Table S2). For our remotely sensed analysis,
312 Green NDVI, NDSVI and their interaction explained 35% of the variation in pre-fire TVC
313 (Appendix S1: Table S2). This predicted TVC had a positive relationship with burn severity
314 ($p \ll 0.01$, $R^2 = .42$, Figure 4b, Appendix S1: Table S2).

315 The majority of seeds that germinated in the greenhouse were the two most common grass
316 species, *P. secunda* and *B. tectorum* (Appendix S1: Table S3, Fig. S2). Eight dicot species
317 were found in more than one location, and these 10 prevalent species are those that were
318 used in our JSDM. Burned sites had an average of 34 ± 32 total seeds in the top 2 cm, and
319 12 ± 14 in the bottom 4 cm. Unburned sites had an average of 299 ± 170 in the top 2 cm
320 and 59 ± 29 in the bottom 4 cm (Appendix S1: Fig. S3). For H2, the JSDM converged
321 well (Appendix S1: Fig S4). Gelman diagnostics were all very close to 1 and the effective
322 sample size centered on 4,000, which indicated good model convergence. Elevation had the
323 strongest effects on individual species occurrence and explained the most variance on average
324 (36%). Burn severity explained 23% of the variance on average and was supported at the 95%
325 level for 5 species (Appendix S1: Fig S2b). For the introduced species, the predictions along
326 a gradient of burn severity were positive for *B. tectorum*, *Sisymbrium altissimum* L. and
327 *Lepidium perfoliatum* L., and negative for *Ceratocephala testiculata* and *Alyssum desertorum*
328 Stapf (Figure 4e). For native species, the effect of burn severity on occurrence was positive
329 for *A. tridentata*, but the mean predictions were still low, never rising above 50%. It was
330 neutral for *P. secunda* and negative for the remaining species. Testing H2a revealed a positive
331 relationship between pre-fire aboveground species diversity and pre-fire fuel connectivity in
332 the single model, and neutral relationships in the path model, and so we felt it was reasonable
333 to rule out pre-fire fuel connectivity as a confounding factor for H2. Testing H2b showed a
334 negative relationship, allowing us to rule out the idea that both pre-fire connectivity and
335 post-fire seed bank composition were simply a function of pre-fire annual grass cover.

336 For H3, we found that, after accounting for elevation, pre-fire aboveground richness, and
337 the number of *P. secunda* seeds, the number of *B. tectorum* seeds in the post-fire seedbank
338 was positively associated with the fuel connectivity in 2019 ($\beta = 0.54$, $p = 0.01$, $\text{Adj } R^2 =$
339 0.75 , Figure 4c, Appendix S1: Table S2). For H4 the most parsimonious model ($\text{Adj } R^2 =$
340 0.89 , Appendix S1: Table S2) had elevation, aspect, fuel connectivity and an interaction
341 between elevation and fuel connectivity as predictors of aboveground Shannon-Weaver alpha

342 diversity. Fuel connectivity was negatively associated with Shannon-Weaver diversity ($\beta =$
343 -0.28 , $p=0.004$, Figure 4d).

344 4. Discussion

345 Here we document how changes in ecosystem structure brought on by invasion can lead
346 to cascading effects on ecosystem function and composition via changes in the disturbance
347 regime. It has already been shown that *B. tectorum* invasion increases fire frequency (Balch
348 et al. 2013), and is indicative of a grass-fire cycle. However, an understanding of the positive
349 feedback mechanisms that link *B. tectorum* invasion success to fire occurrence is required
350 to infer the long-term persistence of such a cycle. The interaction between burn severity
351 and seed bank composition documented here may explain that link. Prior work has shown
352 that annual grass invasion increases fuel connectivity by filling in shrub interspaces with a
353 contiguous bed of fine fuels (Davies and Nafus 2013). This change in the spatial distribution
354 of fine fuels has been associated with larger and more frequent fires (Balch et al. 2013).
355 Here, we found higher fuel connectivity (via TVC) increased burn severity (H1, Figure 4b).
356 Higher burn severity was associated with an increased occurrence of introduced annuals in
357 the post-fire seedbank and a decreased occurrence of native plants with the exception of
358 *A. tridentata* (H2, Figure 4e), but the gains of *A. tridentata* would likely not be enough to
359 counter the gains of *B. tectorum*, especially after a few years of annual grass reproduction
360 and population growth without similar gains for the shrubs (Shriver et al. 2019). Finally,
361 greater abundance of *B. tectorum* seeds in the post-fire seedbank resulted in higher post-fire
362 fuel connectivity (H3, Figure 4c). In addition, we found evidence that high post-fire fuel
363 connectivity was associated with lower aboveground diversity (H4, Figure 4d). This suggests
364 that during inter-fire intervals, there may be additional mechanisms (e.g. competition, altered
365 ecohydrology) maintaining the post-fire, annual grass-dominated species assemblage.

366 The difference in species composition before and after fire explains an apparent contradiction

367 in results between H2a (positive to neutral relationship between pre-fire fuel connectivity and
368 diversity) and H4 (negative relationship between post-fire fuel connectivity and diversity).
369 Most site locations had mature canopies of native shrubs with the inter-shrub space occupied
370 mostly by native bunchgrasses and forbs, with no fire occurrence since 1984. Even in
371 locations with high annual grass cover between shrubs, shrubs provide ecosystem structural
372 heterogeneity and islands of fertility (Doescher, Miller, and Winward 1984; Bechtold and
373 Inouye 2007), and perennial natives that may have been established before invasion have deep
374 roots established that allow for the avoidance of competition for water with shallow-rooted
375 annuals (Gibbens and Lenz 2001; Ottaviani et al. 2020). This may provide enough niche
376 compartmentalization to allow native plants to persist in spite of the invasion prior to fire
377 occurrence. Three years after fire, almost all of the sites were dominated by introduced
378 annuals, and lacked any structural heterogeneity (Appendix S1, Fig. S6c). Thus native
379 plants may have been able to persist via niche compartmentalization after the initial invasion,
380 but fire burned away most of the seeds (Appendix S1, Fig. S3, S7) and removed all of the
381 structural benefits, and microclimatic refugia that shrub cover provides. In this clean slate
382 post-fire environment, the altered species composition of the seedbank and superior post-fire
383 dispersal of *B. tectorum* (Monty, Brown, and Johnston 2013) allow the process of interspecific
384 competition to be dominant (Schlaepfer, Lauenroth, and Bradford 2014).

385 *Contrasts among forests and shrublands as it pertains to remote sensing*

386 Burn severity metrics like dNBR were conceived of in the context of forested ecosystems, and
387 calibrated using the composite burn index (Key and Benson 1999), tree mortality, and percent
388 change in tree canopy cover (J. D. Miller et al. 2009). It is unclear how well these metrics
389 carry over to shrubland systems. We recorded qualitative observations of burn severity while
390 we were sampling, mainly to ensure that we sampled a range of severities, and the dNBR
391 we used appears to correspond with our observations. In areas where the space between
392 shrubs was well-connected by fine fuels (Figure 1 a-c) the burn severity was higher, and the
393 shrubs had completely burned throughout the root system, leaving only a hole in the ground

394 filled with ashes as evidence of their prior presence. In these areas the entirety of the soil
395 surface—underneath shrub canopy and in canopy interspaces—was consumed by fire, and
396 there was little evidence of remaining litter or biological soil crust. Areas with lower fuel
397 connectivity had lower burn severity (Figure 1 d-f). Here, shrubs were usually consumed
398 only to the stumps, and sometimes left standing and charred, destined for mortality. In
399 these areas the soil surface often still had biological soil crust, partially consumed litter
400 (R. O. Jones et al. 2015) and unconsumed annual and perennial grass bases. The manual
401 severity classification provided by MTBS had exclusively low and medium severity, but our
402 observations of essentially complete consumption of plant and litter tissues and very few
403 unburned patches suggested that these should have been mostly medium and high severity.
404 This discrepancy was not unexpected, as the ordinal burn severity classifications produced by
405 MTBS are known to be flawed for research use (Kolden, Smith, and Abatzoglou 2015).

406 Spectral reflectance has long been used to characterize ecosystem structure, including wildfire
407 fuels. Unique signatures of remotely-sensed spectral reflectance are typically matched to
408 categorical fuel classifications (CFCs), which describe the physiognomy of vegetation and
409 its potential to support various fire behavior (Ottmar et al. 2007). While different CFCs
410 can provide a general understanding of fuel amount and connectivity, recent efforts using
411 data with finer spatial and spectral resolution may improve fuel classification with more
412 continuous, multi-dimensional measurements (Stavros et al. 2018). The continuous measure
413 of NDVI in western U.S. coniferous forests is a proxy for live fuel biomass, which likely
414 explains its positive association with wildfire severity (Sean A. Parks et al. 2018; Koontz et al.
415 2020). NDVI also correlates with vegetation cover in these forested systems, and so greater
416 crown connectivity may also explain the NDVI/severity relationship at local scales. When
417 using a more direct NDVI-derived measure of vegetation connectivity in Sierra Nevada yellow
418 pine/mixed-conifer, Koontz et al. (2020) found that greater variability in forest structure,
419 decreased the probability of high-severity fire, likely due to decreased fuel connectivity (i.e.,
420 live tree canopies in the yellow pine/mixed-conifer forest). Here, we arrived at a combination

421 of NDVI and NDSVI to describe the fuel connectivity of the annual grass invaded Great
422 Basin sagebrush community to better reflect key differences in the physiognomies of forest
423 and arid shrublands. In sagebrush shrublands, the fuel that contributes to large wildfires is a
424 mixture of evergreen shrubs interspersed with herbaceous plants that remain green for only a
425 portion of the growing season, and then become dry and straw-colored. Thus, both the live
426 and dead fuel need to be taken into account in remote measurements of fuel connectivity for
427 this system.

428 *Management implications*

429 These results demonstrate that the strength of the grass-fire cycle in this system is controlled
430 by measurable fire properties and ecosystem structural components. We found that annual
431 grass cover was not the single variable that explained burn severity and fuel connectivity
432 (Appendix S1, Fig S5). Rather, it was the contribution of annual grass cover to the total
433 connectivity of the system (Figure 2). The most important areas to prioritize for management
434 interventions could paradoxically be areas with relatively low levels of annual grass cover that
435 join previously disconnected vegetation. Land managers may be able to increase their chances
436 of restoration success by using existing methods or developing novel ones that manipulate
437 these components to weaken or even break the positive feedback cycle. This work provides
438 further evidence that the post-fire annual grassland is a system where the degraded state
439 represents an alternative species assemblage from that of the restoration target. Because the
440 propagules of the original assemblage are no longer present, methods that rely on natural
441 succession may not be sufficient (Suding, Gross, and Houseman 2004). One-off seeding
442 treatments have a low probability of success (Pyke et al. 2020; Arkle et al. 2022), and more
443 labor-intensive methods involving site preparation (Farrell, Fehmi, and Gornish 2021), seed
444 coating and priming (Pedrini et al. 2020), as well as planting live plants (Pyke et al. 2020)
445 may improve the probability of success, as will prioritizing efforts in cooler, wetter years
446 (Bradford et al. 2018; Hardegree et al. 2018; Shriver et al. 2018). Estimating burn severity
447 using satellite imagery may be used in conjunction with site suitability and climate forecasts

448 to help land managers identify areas with a greater likelihood of successful seeding. Our
449 results highlight the importance of prioritizing the preservation of existing native shrub cover
450 and in particular policies that encourage land managers to maximize the preservation of
451 unburned patches within the fire perimeter during the suppression of wildfires in this system
452 (Steenvoorden et al. 2019), as these are the primary sources of native propagules and act as
453 nurse plants (Arkle et al. 2022). In many areas, conditions are now or will in the near future
454 be unsuitable for sagebrush due to annual grass dominance and increases in aridity (Shriver
455 et al. 2019). In these areas it may still be feasible to restore the system’s ability to sequester
456 carbon by planting other native woody species that are more drought tolerant and resilient
457 against fire.

458 Livestock grazing can reduce fuel connectivity in uninvaded sagebrush (Davies et al. 2010).
459 At the same time, livestock grazing can decrease the resistance to invasion by *B. tectorum* via
460 negative effects on biological soil crust (BSC) (Condon and Pyke 2018), and can reduce the
461 survival of *Artemisia* seedlings that are not protected by shrub canopies (Owens and Norton
462 1992). Targeted spring grazing in annual grass monocultures may reduce fuel connectivity
463 and alleviate fire risk. Post-fire grazing may help reduce *B. tectorum* cover, but it may
464 also exacerbate the problem by introducing *B. tectorum* in uninvaded sites (Williamson et
465 al. 2019) or increasing the already superior post-fire dispersal of *B. tectorum* seeds (Monty,
466 Brown, and Johnston 2013). Management interventions should be specifically tailored each
467 year to the conditions of a given site, and focused on native plant restoration.

468 Herbaceous cover in these dryland systems has high interannual variability (Mahood et al.
469 2022). Because the components of ecosystem structure and disturbance severity in positive
470 feedback cycle described here are continuous mechanistic variables, it may be possible to
471 develop theoretical models (*sensu* (Archibald, Staver, and Levin 2012)) to estimate the
472 threshold of vegetation cover that will lead to high burn severity. These can then be applied
473 in conjunction with near real time fuel loading forecasts (M. O. Jones et al. 2021) to identify
474 areas that are vulnerable to high severity fire, which can be used by land managers to take

475 preemptive measures in high value areas.

476 *Global environmental change implications*

477 Understanding how different facets of global environmental change create multiple mechanisms
478 that act in concert to drive ecosystem transformation will provide important insights about
479 ecosystem change from regional to global scales. The system studied here has at least four
480 external processes that may influence the positive feedback we documented. First, land use
481 change via livestock grazing facilitates invasion (Ponzetti, Mccune, and Pyke 2007; Williamson
482 et al. 2019). Second, the introduction of exotic grasses increases fuel connectivity (Davies
483 and Nafus 2013), affects burn severity. Third, increasing temperatures due to climate change
484 increase burn severity in forests (S. A. Parks and Abatzoglou 2020). We expect this to be
485 true for shrublands, and is an important area for future research. Increasing temperatures
486 simultaneously decrease seed viability and seedling survival (Schlaepfer, Lauenroth, and
487 Bradford 2014; Enright et al. 2015). Fourth, CO₂ enrichment may preferentially enhance
488 biomass (i.e. higher fuel connectivity) and seed production of annual grass species (Smith
489 et al. 2000; Nagel et al. 2004). All four of these external drivers are globally ubiquitous
490 consequences of global change.

491 An ecosystem “state” is the product of countless endogenous interactions. The grass-fire
492 cycle studied here is strengthened through providing fitness benefits to the introduced annual
493 grasses via at least three reinforcing processes. First, we document how it changes the
494 composition of the seedbank. Second, introduced annual grasses competitively exclude native
495 plants. Third, the dominance of introduced annual grasses initiates ecohydrological feedbacks
496 to create a warmer, drier microclimate (Turnbull et al. 2012). It is possible that some
497 of these feedbacks are idiosyncratic to the system being studied, while others may reflect
498 fundamental properties of ecosystem function that change when a system is converted from
499 being dominated by deep-rooted woody plants to being dominated by annual herbaceous
500 plants (Kitzberger et al. 2016). At least 13 grass species initiate self-reinforcing feedbacks

501 with fire in the U.S. alone (Fusco et al. 2019; Tortorelli, Krawchuk, and Kerns 2020). There
502 are many more fire-inducing grass invasions worldwide, with documented cases in Australia
503 (G. Miller et al. 2010), Brazil (Rossi et al. 2014) and South Africa (Milton 2004). The
504 conversion of forests and shrublands to grasslands may have consequences relevant to the
505 global carbon cycle, especially when ecosystems dominated by deep-rooted plants that store
506 carbon belowground are replaced by shallow-rooted ecosystems that lose carbon to grazing
507 and fire (Kerns et al. 2020; Mahood et al. 2022).

508 **Acknowledgements**

509 We thank Abdelhakim Farid, Julia Lopez, Dylan Murphy and C. Nick Whittemore for help
510 in the field and in the greenhouse, and Lindsay P. Chiquoine, Thomas T. Veblen and two
511 anonymous reviewers for constructive feedback that greatly improved the manuscript. We
512 are grateful to everyone in the Winnemucca office of the Bureau of Land Management, the
513 Central Nevada Interagency Dispatch Center, and CU Boulder’s Ecology Evolution and
514 Biology Greenhouse. This project was funded in part by the CU Boulder Undergraduate
515 Research Opportunities Program and the Geography department’s Adam Kolff Memorial
516 Graduate Research Grant.

517 **References**

- 518 Aleman, J. C., A. Fayolle, C. Favier, A. C. Staver, K. G. Dexter, C. M. Ryan, A. F.
519 Azihou, et al. 2020. “Floristic Evidence for Alternative Biome States in Tropical
520 Africa.” *Proceedings of the National Academy of Sciences* 117 (45): 28183–90. <https://doi.org/10.1073/pnas.2011515117>.
521
- 522 Allan, Eric, Pete Manning, Fabian Alt, Julia Binkenstein, Stefan Blaser, Nico Blüthgen,
523 Stefan Böhm, et al. 2015. “Land Use Intensification Alters Ecosystem Multifunctionality
524 via Loss of Biodiversity and Changes to Functional Composition.” *Ecology Letters* 18 (8):

525 834–43. <https://doi.org/10.1111/ele.12469>.

526 Archibald, Sally, A. Carla Staver, and Simon A. Levin. 2012. “Evolution of Human-Driven
527 Fire Regimes in Africa.” *Proceedings of the National Academy of Sciences* 109 (3): 847–52.
528 <https://doi.org/10.1073/pnas.1118648109>.

529 Arkle, Robert S., David S. Pilliod, Matthew J. Germino, Michelle I. Jeffries, and Justin
530 L. Welty. 2022. “Reestablishing a Foundational Species: Limitations on Post-wildfire
531 Sagebrush Seedling Establishment.” *Ecosphere* 13 (8). <https://doi.org/10.1002/ecs2.4195>.

532 Balch, Jennifer K., Bethany A. Bradley, Carla M. D’Antonio, and José Gómez-Dans. 2013.
533 “Introduced annual grass increases regional fire activity across the arid western USA
534 (1980-2009).” *Global Change Biology* 19 (1): 173–83. <https://doi.org/10.1111/gcb.12046>.

535 Bechtold, H. A., and R. S. Inouye. 2007. “Distribution of carbon and nitrogen in sage-
536 brush steppe after six years of nitrogen addition and shrub removal.” *Journal of Arid
537 Environments* 71 (1): 122–32. <https://doi.org/10.1016/j.jaridenv.2007.02.004>.

538 Beckstead, Julie, Laura E. Street, Susan E. Meyer, and Phil S. Allen. 2011. “Fire effects on
539 the cheatgrass seed bank pathogen *Pyrenophora semeniperda*.” *Rangeland Ecology and
540 Management* 64 (2): 148–57. <https://doi.org/10.2111/REM-D-10-00052.1>.

541 Bond, William J., and Jeremy J. Midgley. 1995. “Kill Thy Neighbour: An Individualistic
542 Argument for the Evolution of Flammability.” *Oikos* 73 (1): 79. [https://doi.org/10.2307/
543 3545728](https://doi.org/10.2307/3545728).

544 Boudell, JA, SO Link, and JR Johansen. 2002. “Effect of soil microtopography on seed
545 bank distribution in the shrub-steppe.” *Western North American Naturalist* 62 (1): 14–24.
546 <https://doi.org/10.2307/41717153>.

547 Bradford, John B, Julio L Betancourt, Bradley J Butterfield, Seth M Munson, and Troy E
548 Wood. 2018. “Anticipatory Natural Resource Science and Management for a Changing
549 Future.” *Frontiers in Ecology and the Environment* 16 (5): 295–303. [https://doi.org/10.
550 1002/fee.1806](https://doi.org/10.1002/fee.1806).

551 Brooks, Matthew L., Carla M. D’Antonio, David M. Richardson, James B. Grace, Jon E.

552 Keeley, Joseph M. DiTomaso, Richard J. Hobbs, Mike Pellant, and David Pyke. 2004.
553 “Effects of Invasive Alien Plants on Fire Regimes.” *BioScience* 54 (7): 677–88.

554 Bukowski, Beth, and William L. Baker. 2013. “Historical fire regimes, reconstructed from
555 land-survey data, led to complexity and fluctuation in sagebrush landscapes.” *Ecological*
556 *Applications* 23 (3): 546–64.

557 Commission for Environmental Cooperation. 2006. “Ecological regions of North America –
558 Levels I, II, and III: Montreal, Quebec, Canada, Commission for Environmental Coopera-
559 tion, scale 1:10,000,000.” <https://www.epa.gov/eco-research/ecoregions-north-america>.

560 Condon, Lea A., and David A. Pyke. 2018. “Fire and Grazing Influence Site Resistance to
561 *Bromus tectorum* Through Their Effects on Shrub, Bunchgrass and Biocrust Communities
562 in the Great Basin (USA).” *Ecosystems* 21 (7): 1416–31. [https://doi.org/10.1007/s10021-](https://doi.org/10.1007/s10021-018-0230-8)
563 [018-0230-8](https://doi.org/10.1007/s10021-018-0230-8).

564 D’Antonio, Carla M., and Peter M. Vitousek. 1992. “Biological invasions by exotic grasses,
565 the grass/fire cycle, and global change.” *Annual Review of Ecological Systems* 23: 63–87.

566 Davies, Kirk W., Jonathan D. Bates, Tony J. Svejcar, and Chad S. Boyd. 2010. “Effects
567 of long-term livestock grazing on fuel characteristics in rangelands: An example from
568 the sagebrush steppe.” *Rangeland Ecology and Management* 63 (6): 662–69. <https://doi.org/10.2111/REM-D-10-00006.1>.

570 Davies, Kirk W., Chad S. Boyd, Jeffrey L. Beck, Jon D. Bates, Tony J. Svejcar, and
571 Michael A. Gregg. 2011. “Saving the sagebrush sea: An ecosystem conservation plan
572 for big sagebrush plant communities.” *Biological Conservation* 144 (11): 2573–84. <https://doi.org/10.1016/j.biocon.2011.07.016>.

574 Davies, Kirk W., and Aleta M. Nafus. 2013. “Exotic annual grass invasion alters fuel amounts,
575 continuity and moisture content.” *International Journal of Wildland Fire* 22 (3): 353–58.
576 <https://doi.org/10.1071/WF11161>.

577 Davis, Kimberley T., Solomon Z. Dobrowski, Philip E. Higuera, Zachary A. Holden, Thomas
578 T. Veblen, Monica T. Rother, Sean A. Parks, Anna Sala, and Marco P. Maneta. 2019.

579 “Wildfires and Climate Change Push Low-Elevation Forests Across a Critical Climate
580 Threshold for Tree Regeneration.” *Proceedings of the National Academy of Sciences*,
581 201815107. <https://doi.org/10.1073/pnas.1815107116>.

582 Davis, Kimberley T., Philip E. Higuera, and Anna Sala. 2018. “Anticipating Fire-mediated
583 Impacts of Climate Change Using a Demographic Framework.” Edited by Charles Fox.
584 *Functional Ecology* 32 (7): 1729–45. <https://doi.org/10.1111/1365-2435.13132>.

585 Doescher, Paul S., Richard F. Miller, and Alma H. Winward. 1984. “Soil Chemical Patterns
586 under Eastern Oregon Plant Communities Dominated by Big Sagebrush.” <https://doi.org/10.2136/sssaj1984.03615995004800030038x>.

587

588 Eckert, Richard E., Frederick F. Peterson, Michael S. Meurisse, and L. Stephens. 1986.
589 “Effects of Soil-Surface Morphology on Emergence and Survival of Seedlings in Big
590 Sagebrush Communities.” *Journal of Range Management* 39 (5): 414–20. <http://www.jstor.org/stable/3899441>.

591

592 Eidenshink, Jeff, Brian Schwind, Ken Brewer, Zhi-liang Zhu, Brad Quayle, and Stephen
593 Howard. 2007. “A Project for Monitoring Trends in Burn Severity.” *Fire Ecology* 3 (1):
594 3–21. <https://doi.org/10.4996/fireecology.0301003>.

595 Elzinga, Caryl L, Daniel W Salzer, and John W Willoughby. 1998. “Measuring & Monitoring
596 Plant Populations.” BLM Technical Reference 1730-1. Bureau of Land Management.

597 Enright, Neal J., Joseph B. Fontaine, David M. J. S. Bowman, Ross A. Bradstock, and
598 Richard J. Williams. 2015. “Interval Squeeze: Altered Fire Regimes and Demographic
599 Responses Interact to Threaten Woody Species Persistence as Climate Changes.” *Frontiers*
600 *in Ecology and the Environment* 13 (5): 265–72. <https://doi.org/10.1890/140231>.

601 Eskelinen, Anu, Kelly Gravuer, W Stanley Harpole, Susan Harrison, Risto Virtanen, and
602 Yann Hautier. 2020. “Resource-enhancing Global Changes Drive a Whole-ecosystem Shift
603 to Faster Cycling but Decrease Diversity.” *Ecology* 101 (12). [https://doi.org/10.1002/ecy.](https://doi.org/10.1002/ecy.3178)
604 [3178](https://doi.org/10.1002/ecy.3178).

605 Farrell, Hannah L., Jeffrey S. Fehmi, and Elise S. Gornish. 2021. “The Effects of Site

606 Preparation Equal Those of Seeding at a Dryland Restoration Site: 6 Years of Plant
607 Community Development.” *Restoration Ecology* 29 (6). <https://doi.org/10.1111/rec.13482>.

608 Fenesi, Annamária, Sandra Saura-Mas, Robert R. Blank, Anita Kozma, Beáta-Magdolna
609 Lózer, and Eszter Ruprecht. 2016. “Enhanced Fire-Related Traits May Contribute to
610 the Invasiveness of Downy Brome (*Bromus Tectorum*).” *Invasive Plant Science and*
611 *Management* 9 (3): 182–94. <https://doi.org/10.1614/IPSM-D-16-00006.1>.

612 Fusco, Emily J., John T. Finn, Jennifer K. Balch, R. Chelsea Nagy, and Bethany A. Bradley.
613 2019. “Invasive Grasses Increase Fire Occurrence and Frequency Across US Ecoregions.”
614 *Proceedings of the National Academy of Sciences* 116 (47): 23594–99. [https://doi.org/10.](https://doi.org/10.1073/pnas.1908253116)
615 [1073/pnas.1908253116](https://doi.org/10.1073/pnas.1908253116).

616 Gagnon, Paul R., Heather A. Passmore, Matthew Slocum, Jonathan A. Myers, Kyle E.
617 Harms, William J. Platt, and C. E. Timothy Paine. 2015. “Fuels and fires influence
618 vegetation via above- and belowground pathways in a high-diversity plant community.”
619 *Journal of Ecology* 103 (4): 1009–19. <https://doi.org/10.1111/1365-2745.12421>.

620 Gelman, Andrew, Donald B Rubin, et al. 1992. “Inference from Iterative Simulation Using
621 Multiple Sequences.” *Statistical Science* 7 (4): 457–72.

622 Germino, Matthew J., Jeanne C. Chambers, and Cynthia S. Brown. 2016. *Exotic Brome-*
623 *Grasses in Arid and Semiarid Ecosystems of the Western US Causes, Consequences, and*
624 *Management Implications*. <http://www.springer.com/series/412>.

625 Gibbens, Robert P., and James M. Lenz. 2001. “Root systems of some Chihuahuan Desert
626 plants.” *Journal of Arid Environments* 49: 221–63.

627 Gill, Nathan S, Monica G Turner, Carissa D Brown, Sydney I Glassman, Sandra L Haire,
628 Winslow D Hansen, Elizabeth R Pansing, Samuel B St Clair, and Diana F Tomback. 2022.
629 “Limitations to Propagule Dispersal Will Constrain Postfire Recovery of Plants and Fungi
630 in Western Coniferous Forests.” *BioScience*, January, biab139. [https://doi.org/10.1093/](https://doi.org/10.1093/biosci/biab139)
631 [biosci/biab139](https://doi.org/10.1093/biosci/biab139).

632 Hardegree, Stuart P., John T. Abatzoglou, Mark W. Brunson, Matthew J. Germino, Katherine

- 633 C. Hegewisch, Corey A. Moffet, David S. Pilliod, Bruce A. Roundy, Alex R. Boehm,
634 and Gwendwr R. Meredith. 2018. “Weather-Centric Rangeland Revegetation Planning.”
635 *Rangeland Ecology & Management* 71 (1): 1–11. [https://doi.org/10.1016/j.rama.2017.07.](https://doi.org/10.1016/j.rama.2017.07.003)
636 [003](https://doi.org/10.1016/j.rama.2017.07.003).
- 637 Hassan, M. A., and N. E. West. 1986. “Dynamics of Soil Seed Pools in Burned and Unburned
638 Sagebrush Semi-Deserts.” *Ecology* 67 (1): 269–72.
- 639 Heerdt, G. N. J. Ter, G. L. Verweij, R. M. Bekker, and J. P. Bakker. 1996. “An Improved
640 Method for Seed-Bank Analysis: Seedling Emergence After Removing the Soil by Sieving.”
641 *Functional Ecology* 10 (1): 144. <https://doi.org/10.2307/2390273>.
- 642 Heydari, Mehdi, Reza Omidipour, Mehdi Abedi, and Carol Baskin. 2017. “Effects of
643 fire disturbance on alpha and beta diversity and on beta diversity components of soil
644 seed banks and aboveground vegetation.” *Plant Ecology and Evolution* 150 (3): 247–56.
645 <https://doi.org/10.5091/plecevo.2017.1344>.
- 646 Hirota, Marina, Milena Holmgren, Egbert H. Van Nes, and Marten Scheffer. 2011. “Global
647 resilience of tropical forest and savanna to critical transitions.” *Science* 334 (6053): 232–35.
648 <https://doi.org/10.1126/science.1210657>.
- 649 Humphrey, L David, and Eugene W Schupp. 2001. “Seed banks of *Bromus tectorum*-
650 dominated communities in the Great Basin.” *Western North American Naturalist* 61 (1):
651 85–92. <https://doi.org/10.2307/41717080>.
- 652 Jones, Matthew O., Nathaniel P. Robinson, David E. Naugle, Jeremy D. Maestas, Matthew
653 C. Reeves, Robert W. Lankston, and Brady W. Allred. 2021. “Annual and 16-Day
654 Rangeland Production Estimates for the Western United States.” *Rangeland Ecology &*
655 *Management* 77 (July): 112–17. <https://doi.org/10.1016/j.rama.2021.04.003>.
- 656 Jones, Rachel O., Jeanne C. Chambers, David I. Board, Dale W. Johnson, and Robert R.
657 Blank. 2015. “The role of resource limitation in restoration of sagebrush ecosystems
658 dominated by cheatgrass (*Bromus tectorum*).” *Ecosphere* 6 (7): 1–21.
- 659 Keddy, Paul A, and Daniel C Laughlin. 2021. *A Framework for Community Ecology: Species*

660 *Pools, Filters and Traits*. Cambridge University Press.

661 Keeley, Jon E. 2009. “Fire intensity, fire severity and burn severity: A brief review and
662 suggested usage.” *International Journal of Wildland Fire* 18 (1): 116–26. [https://doi.org/
663 10.1071/WF07049](https://doi.org/10.1071/WF07049).

664 Keeley, Jon E., Juli G. Pausas, Philip W. Rundel, William J. Bond, and Ross A. Bradstock.
665 2011. “Fire as an evolutionary pressure shaping plant traits.” *Trends in Plant Science* 16
666 (8): 406–11. <https://doi.org/10.1016/j.tplants.2011.04.002>.

667 Kerns, Becky K., Claire Tortorelli, Michelle A. Day, Ty Nietupski, Ana M. G. Barros, John
668 B. Kim, and Meg A. Krawchuk. 2020. “Invasive grasses: A new perfect storm for
669 forested ecosystems?” *Forest Ecology and Management* 463 (November 2019): 117985.
670 <https://doi.org/10.1016/j.foreco.2020.117985>.

671 Key, Carl H, and Nathan C Benson. 1999. “The Composite Burn Index (CBI): Field Rating
672 of Burn Severity.” *USGS, NRMSC Research,[online] Available: Http://Nrmsc. Usgs.
673 Gov/Research/Cbi. Htm [3/14/2006]*.

674 Kimura, Hideo, and Shiro Tsuyuzaki. 2011. “Fire severity affects vegetation and seed bank
675 in a wetland.” *Applied Vegetation Science* 14 (3): 350–57. [https://doi.org/10.1111/j.1654-
676 109X.2011.01126.x](https://doi.org/10.1111/j.1654-109X.2011.01126.x).

677 Kitzberger, Thomas, G. L. W. Perry, J. Paritsis, J. H. Gowda, A. J. Tepley, A. Holz, and T.
678 T. Veblen. 2016. “Fire–vegetation feedbacks and alternative states: common mechanisms
679 of temperate forest vulnerability to fire in southern South America and New Zealand.”
680 *New Zealand Journal of Botany* 54 (2): 247–72. [https://doi.org/10.1080/0028825X.2016.
681 1151903](https://doi.org/10.1080/0028825X.2016.1151903).

682 Kolden, Crystal A, Alistair M S Smith, and John T. Abatzoglou. 2015. “Limitations and
683 utilisation of Monitoring Trends in Burn Severity products for assessing wildfire severity
684 in the USA.” *International Journal of Wildland Fire* 24: 1023–28.

685 Koontz, Michael J, Malcolm P North, Chhaya M Werner, Stephen E Fick, and Andrew M
686 Latimer. 2020. “Local Forest Structure Variability Increases Resilience to Wildfire in Dry

687 Western US Coniferous Forests.” *Ecology Letters* 23 (3): 483–94.

688 Lipoma, M. Lucrecia, Guillermo Funes, and Sandra Díaz. 2018. “Fire effects on the soil
689 seed bank and post-fire resilience of a semi-arid shrubland in central Argentina.” *Austral*
690 *Ecology* 43 (1): 46–55. <https://doi.org/10.1111/aec.12533>.

691 Liyanage, Ganesh S., and Mark K. J. Ooi. 2017. “Do dormancy-breaking temperature
692 thresholds change as seeds age in the soil seed bank?” *Seed Science Research* 27 (1): 1–11.
693 <https://doi.org/10.1017/S0960258516000271>.

694 Mahood, Adam L., and Jennifer K. Balch. 2019. “Repeated fires reduce plant diversity
695 in low-elevation Wyoming big sagebrush ecosystems (1984 – 2014).” *Ecosphere* 10 (2):
696 e02591. <https://doi.org/10.1002/ecs2.2591>.

697 Mahood, Adam L., Rachel O. Jones, David I. Board, Jennifer K. Balch, and Jeanne C.
698 Chambers. 2022. “Interannual Climate Variability Mediates Changes in Carbon and
699 Nitrogen Pools Caused by Annual Grass Invasion in a Semiarid Shrubland.” *Global Change*
700 *Biology* 28 (1): 267–84. <https://doi.org/10.1111/gcb.15921>.

701 Maia, P., J. G. Pausas, V. Arcenegui, C. Guerrero, A. Pérez-Bejarano, J. Mataix-Solera, M.
702 E. T. Varela, I. Fernandes, E. T. Pedrosa, and J. J. Keizer. 2012. “Wildfire effects on the
703 soil seed bank of a maritime pine stand - The importance of fire severity.” *Geoderma* 191:
704 80–88. <https://doi.org/10.1016/j.geoderma.2012.02.001>.

705 Meyer, Susan E, Stephen B Monsen, and E Durant Mearthur. 2013. “Germination Response
706 of *Artemisia tridentata* (Asteraceae) to Light and Chill: Patterns of Between-Population
707 Variation.” *Botanical Gazette* 151 (2): 176–83.

708 Miller, Georgia, Margaret Friedel, Paul Adam, and Vanessa Chewings. 2010. “Ecological
709 impacts of buffel grass (*Cenchrus ciliaris* L.) invasion in central Australia does field
710 evidence support a fire-invasion feedback?” *The Rangeland Journal* 32 (4): 353–65.
711 <https://doi.org/10.1071/RJ09076>.

712 Miller, Jay D., Eric E. Knapp, Carl H. Key, Carl N. Skinner, Clint J. Isbell, R. Max Creasy,
713 and Joseph W. Sherlock. 2009. “Calibration and validation of the relative differenced

714 Normalized Burn Ratio (RdNBR) to three measures of fire severity in the Sierra Nevada
715 and Klamath Mountains, California, USA.” *Remote Sensing of Environment* 113 (3):
716 645–56. <https://doi.org/10.1016/j.rse.2008.11.009>.

717 Milton, Sue J. 2004. “Grasses as invasive alien plants in South Africa.” *South African Journal*
718 *of Science* 100 (1-2): 69–75.

719 Monty, Arnaud, Cynthia S. Brown, and Danielle B. Johnston. 2013. “Fire promotes downy
720 brome (*Bromus tectorum* L.) seed dispersal.” *Biological Invasions* 15 (5): 1113–23.
721 <https://doi.org/10.1007/s10530-012-0355-1>.

722 Morgan, P., and L. F. Neuenschwander. 1988. “Seed-Bank Contributions to Regeneration of
723 Shrub Species After Clear-Cutting and Burning.” *Canadian Journal of Botany* 66 (1):
724 169–72. <https://doi.org/10.1139/b88-026>.

725 Morris, Lesley R., and Elizabeth A. Leger. 2016. “Secondary Succession in the Sagebrush
726 Semidesert 66 Years After Fire in the Great Basin, USA.” *Natural Areas Journal* 36 (2):
727 187–93. <https://doi.org/10.3375/043.036.0211>.

728 Nagel, Jennifer M., Travis E. Huxman, Kevin L. Griffin, and Stanley D. Smith. 2004. “CO₂
729 enrichment reduces the energetic cost of biomass construction in an invasive desert grass.”
730 *Ecology* 85 (1): 100–106. <https://doi.org/10.1890/02-3005>.

731 Naghipour, Ali Asghar, Hossein Bashari, Seyed Jamaledin Khajeddin, Pejman Tahmasebi,
732 and Majid Iravani. 2016. “Effects of Smoke, Ash and Heat Shock on Seed Germination of
733 Seven Species from Central Zagros Rangelands in the Semi-Arid Region of Iran.” *African*
734 *Journal of Range & Forage Science* 33 (1): 67–71. [https://doi.org/10.2989/10220119.
735 2015.1119194](https://doi.org/10.2989/10220119.2015.1119194).

736 Ottaviani, Gianluigi, Rafael Molina-Venegas, Tristan Charles-Dominique, Stefano Chelli,
737 Giandiego Campetella, Roberto Canullo, and Jitka Klimešová. 2020. “The Neglected
738 Belowground Dimension of Plant Dominance.” *Trends in Ecology and Evolution* 35 (9):
739 763–66. <https://doi.org/10.1016/j.tree.2020.06.006>.

740 Ottmar, Roger D, David V Sandberg, Cynthia L Riccardi, and Susan J Prichard. 2007. “An

741 Overview of the Fuel Characteristic Classification System—Quantifying, Classifying, and
742 Creating Fuelbeds for Resource Planning.” *Canadian Journal of Forest Research* 37 (12):
743 2383–93.

744 Owens, M. K., and B. E. Norton. 1992. “Interactions of Grazing and Plant Protection on
745 Basin Big Sagebrush (*Artemisia tridentata* ssp . *tridentata*) Seedling Survival.” *Journal*
746 *of Range Management* 45 (3): 257–62. <http://www.jstor.org/stable/4002974>.

747 Palmer, Harrison D., Andrew J. Denham, and Mark K. J. Ooi. 2018. “Fire severity drives
748 variation in post-fire recruitment and residual seed bank size of *Acacia* species.” *Plant*
749 *Ecology* 219 (5): 527–37. <https://doi.org/10.1007/s11258-018-0815-5>.

750 Parks, S. A., and J. T. Abatzoglou. 2020. “Warmer and Drier Fire Seasons Contribute to
751 Increases in Area Burned at High Severity in Western US Forests From 1985 to 2017.”
752 *Geophysical Research Letters* 47 (22). <https://doi.org/10.1029/2020GL089858>.

753 Parks, Sean A., Lisa M. Holsinger, Morgan A. Voss, Rachel A. Loehman, and Nathaniel
754 P. Robinson. 2018. “Mean composite fire severity metrics computed with google earth
755 engine offer improved accuracy and expanded mapping potential.” *Remote Sensing* 10
756 (6): 1–15. <https://doi.org/10.3390/rs10060879>.

757 Pedrini, Simone, Alma Balestrazzi, Matthew D. Madsen, Khiraj Bhalsing, Stuart P. Hardegree,
758 Kingsley W. Dixon, and Olga A. Kildisheva. 2020. “Seed Enhancement: Getting Seeds
759 Restoration-ready.” *Restoration Ecology* 28 (S3). <https://doi.org/10.1111/rec.13184>.

760 Petraitis, Peter S., and Roger Earl Latham. 1999. “The importance of scale in testing the
761 origins of alternative community states.” *Ecology* 80 (2): 429–42. [https://doi.org/10.1890/
762 0012-9658\(1999\)080%5B0429:TIOSIT%5D2.0.CO;2](https://doi.org/10.1890/0012-9658(1999)080%5B0429:TIOSIT%5D2.0.CO;2).

763 Ponzetti, J. M., B. Mccune, and David A. Pyke. 2007. “Biotic Soil Crusts in Relation to
764 Topography, Cheatgrass and Fire in the Columbia Basin, Washington.” *The Bryologist* 110
765 (4): 706–22. [https://doi.org/10.1639/0007-2745\(2007\)110%5B706:BSCIRT%5D2.0.CO;2](https://doi.org/10.1639/0007-2745(2007)110%5B706:BSCIRT%5D2.0.CO;2).

766 Pyke, David A., Robert K. Shriver, Robert S. Arkle, David S. Pilliod, Cameron L. Aldridge,
767 Peter S. Coates, Matthew J. Germino, Julie A. Heinrichs, Mark A. Ricca, and Scott E.

768 Shaff. 2020. “Postfire Growth of Seeded and Planted Big Sagebrush—Strategic Designs for
769 Restoring Greater Sage-grouse Nesting Habitat.” *Restoration Ecology* 28 (6): 1495–1504.
770 <https://doi.org/10.1111/rec.13264>.

771 R Core Team. 2020. *R: A Language and Environment for Statistical Computing*. Vienna,
772 Austria: R Foundation for Statistical Computing. <https://www.R-project.org/>.

773 Ratajczak, Zak, Stephen R. Carpenter, Anthony R. Ives, Christopher J. Kucharik, Tanjona
774 Ramiadantsoa, M. Allison Stegner, John W. Williams, Jien Zhang, and Monica G. Turner.
775 2018. “Abrupt Change in Ecological Systems: Inference and Diagnosis.” *Trends in Ecology*
776 *and Evolution* 33 (7): 513–26. <https://doi.org/10.1016/j.tree.2018.04.013>.

777 Rosseel, Yves. 2012. “lavaan: An R Package for Structural Equation Modeling.” *Journal of*
778 *Statistical Software* 48 (2): 1–36. <https://doi.org/10.18637/jss.v048.i02>.

779 Rossi, Rafael Drumond, Carlos Romero Martins, Pedro Lage Viana, Evandro Luís Rodrigues,
780 and José Eugênio Côrtes Figueira. 2014. “Impact of invasion by molasses grass (*Melinis*
781 *minutiflora* P. Beauv.) on native species and on fires in areas of campo-cerrado in Brazil.”
782 *Acta Botanica Brasilica* 28 (4): 631–37. <https://doi.org/10.1590/0102-33062014abb3390>.

783 Rother, Monica T., Thomas T. Veblen, and Luke G. Furman. 2015. “A Field Experiment
784 Informs Expected Patterns of Conifer Regeneration After Disturbance Under Changing
785 Climate Conditions.” *Canadian Journal of Forest Research* 45 (11): 1607–16. <https://doi.org/10.1139/cjfr-2015-0033>.

787 Scheffer, Marten, and Stephen R. Carpenter. 2003. “Catastrophic Regime Shifts in Ecosys-
788 tems: Linking Theory to Observation.” *Trends in Ecology & Evolution* 18 (12): 648–56.
789 <https://doi.org/10.1016/j.tree.2003.09.002>.

790 Scheffer, Marten, Stephen R. Carpenter, Vasilis Dakos, and Egbert H. van Nes. 2015. “Generic
791 Indicators of Ecological Resilience: Inferring the Chance of a Critical Transition.” *Annual*
792 *Review of Ecology, Evolution, and Systematics* 46 (1): 145–67. <https://doi.org/10.1146/annurev-ecolsys-112414-054242>.

794 Scheffer, M, S Carpenter, JA Foley, C Folke, and B Walker. 2001. “Catastrophic Shifts in

795 Ecosystems.” *Nature* 413: 591–96.

796 Schimmel, Johnny, and Anders Granström. 1996. “Fire Severity and Vegetation Response in
797 the Boreal Swedish Forest.” *Ecology* 77 (5): 1436–50.

798 Schlaepfer, Daniel R., William K. Lauenroth, and John B. Bradford. 2014. “Natural
799 Regeneration Processes in Big Sagebrush (*Artemisia tridentata*).” *Rangeland Ecology &
800 Management* 67 (4): 344–57. <https://doi.org/10.2111/REM-D-13-00079.1>.

801 Schröder, Arne, Lennart Persson, and André M. De Roos. 2005. “Direct Experimental
802 Evidence for Alternative Stable States: A Review.” *Oikos* 110 (1): 3–19. [https://doi.org/
803 10.1111/j.0030-1299.2005.13962.x](https://doi.org/10.1111/j.0030-1299.2005.13962.x).

804 Shannon, CE, and W Weaver. 1949. “The Mathematical Theory of Communication. Univer-
805 sity of Illinois Press, Urbana-Champaign, Illinois, USA, 117 p.”

806 Shinneman, Douglas J., and Susan K. McIlroy. 2016. “Identifying key climate and environ-
807 mental factors affecting rates of post-fire big sagebrush (*Artemisia tridentata*) recovery in
808 the northern Columbia Basin, USA.” *International Journal of Wildland Fire* 25: 933–45.
809 <https://doi.org/10.1071/WF16013>.

810 Shriver, Robert K., Caitlin M. Andrews, Robert S. Arkle, David M. Barnard, Michael C.
811 Duniway, Matthew J. Germino, David S. Pilliod, David A. Pyke, Justin L. Welty, and
812 John B. Bradford. 2019. “Transient Population Dynamics Impede Restoration and May
813 Promote Ecosystem Transformation After Disturbance.” Edited by Shaopeng Wang.
814 *Ecology Letters* 22 (9): 1357–66. <https://doi.org/10.1111/ele.13291>.

815 Shriver, Robert K., Caitlin M. Andrews, David S. Pilliod, Robert S. Arkle, Justin L. Welty,
816 Matthew J. Germino, Michael C. Duniway, David A. Pyke, and John B. Bradford.
817 2018. “Adapting Management to a Changing World: Warm Temperatures, Dry Soil,
818 and Interannual Variability Limit Restoration Success of a Dominant Woody Shrub in
819 Temperate Drylands.” *Global Change Biology* 24 (10): 4972–82. [https://doi.org/10.1111/
820 gcb.14374](https://doi.org/10.1111/gcb.14374).

821 Smith, Stanley D., Travis E. Huxman, Stephen F. Zitzer, Therese N. Charlet, David C.

822 Housman, James S. Coleman, Lynn K. Fenstermaker, Jeffrey R. Seemann, and Robert S.
823 Nowak. 2000. “Elevated CO₂ increases productivity and invasive species success in an
824 arid ecosystem.” *Nature* 408 (6808): 79–82. <https://doi.org/10.1038/35040544>.

825 Staal, Arie, Ingo Fetzer, Lan Wang-Erlandsson, Joyce H. C. Bosmans, Stefan C. Dekker,
826 Egbert H. van Nes, Johan Rockström, and Obbe A. Tuinenburg. 2020. “Hysteresis
827 of tropical forests in the 21st century.” *Nature Communications* 11 (1): 1–8. <https://doi.org/10.1038/s41467-020-18728-7>.

829 Staver, A. Carla, Sally Archibald, and Simon A. Levin. 2011. “The global extent and
830 determinants of savanna and forest as alternative biome states.” *Science* 334 (6053):
831 230–32. <https://doi.org/10.1126/science.1210465>.

832 Stavros, E. Natasha, Janice Coen, Birgit Peterson, Harshvardhan Singh, Kama Kennedy,
833 Carlos Ramirez, and David Schimel. 2018. “Use of Imaging Spectroscopy and LIDAR to
834 Characterize Fuels for Fire Behavior Prediction.” *Remote Sensing Applications: Society
835 and Environment* 11: 41–50. <https://doi.org/https://doi.org/10.1016/j.rsase.2018.04.010>.

836 Steenvoorden, Jasper, Arjan J. H. Meddens, Anthony J. Martinez, Lee J. Foster, and W.
837 Daniel Kissling. 2019. “The potential importance of unburned islands as refugia for the
838 persistence of wildlife species in fire-prone ecosystems.” *Ecology and Evolution* 9 (15):
839 8800–8812. <https://doi.org/10.1002/ece3.5432>.

840 Suding, Katharine N., Katherine L. Gross, and Gregory R. Houseman. 2004. “Alternative
841 states and positive feedbacks in restoration ecology.” *Trends in Ecology & Evolution* 19
842 (1): 46–53. <https://doi.org/10.1016/j.tree.2003.10.005>.

843 Tikhonov, Gleb, Otso Ovaskainen, Jari Oksanen, Melinda de Jonge, Oystein Opedal, and
844 Tad Dallas. 2020. *Hmsc: Hierarchical Model of Species Communities*. [https://CRAN.R-
845 project.org/package=Hmsc](https://CRAN.R-project.org/package=Hmsc).

846 Tortorelli, Claire M., Meg A. Krawchuk, and Becky K. Kerns. 2020. “Expanding the
847 invasion footprint: *Venttenata dubia* and relationships to wildfire, environment, and plant
848 communities in the Blue Mountains of the Inland Northwest, USA.” *Applied Vegetation*

849 *Science*, no. May: 1–13. <https://doi.org/10.1111/avsc.12511>.

850 Turnbull, Laura, Bradford P. Wilcox, J. Benlap, S. Ravi, P. D’Odorico, D. Childers, W.
851 Gwenzi, et al. 2012. “Understanding the role of ecohydrological feedbacks in ecosystem
852 state change in drylands.” *Ecohydrology* 5: 174–83. <https://doi.org/10.1002/eco>.

853 U.S. Department of Interior, Bureau of Land Management (BLM). 2018. “BLM AIM
854 TerrADat TerrestrialAIM point.” BLM National Operations Center: BLM. [https://gis.
855 blm.gov/AIMdownload/layerpackages/BLM_AIM_Terrestrial.lpk](https://gis.blm.gov/AIMdownload/layerpackages/BLM_AIM_Terrestrial.lpk).

856 Williamson, Matthew A., Erica Fleishman, Ralph C. Mac Nally, Jeanne C. Chambers,
857 Bethany A. Bradley, David S. Dobkin, David I. Board, et al. 2019. “Fire, livestock
858 grazing, topography, and precipitation affect occurrence and prevalence of cheatgrass
859 (*Bromus tectorum*) in the central Great Basin, USA.” *Biological Invasions* 22 (2): 663–80.
860 <https://doi.org/10.1007/s10530-019-02120-8>.

861 Wright, Boyd R., Peter K. Latz, and A. F. Zuur. 2016. “Fire severity mediates seedling
862 recruitment patterns in slender mulga (*Acacia aptaneura*), a fire-sensitive Australian
863 desert shrub with heat-stimulated germination.” *Plant Ecology* 217 (6): 789–800. [https:
864 //doi.org/10.1007/s11258-015-0550-0](https://doi.org/10.1007/s11258-015-0550-0).

865 **Figure Captions**

866 **Figure 1.** Visual illustration of the relationship between fuel connectivity and burn severity.
867 On the left, panel a shows the inter-shrub space invaded by annual grasses. The photo in
868 panel b was taken in the exact same place two weeks later, days after all of the biomass was
869 consumed by the fire. Panel C is a closeup of the soil surface, showing in more detail how the
870 litter was also almost completely consumed by the fire. On the right, the photos in panels d
871 and e were on opposite sides of a fire line in an area that had minimal annual grass invasion
872 over a broad area, and thus lower fuel connectivity. Note the remaining plants and stumps in
873 panel e and the presence of only partially consumed litter in panel f.

874 **Figure 2.** Sites with little to no shrub cover require high IAG cover to meet the threshold
875 necessary to carry a fire, while sites with higher shrub cover may reach that threshold
876 with much lower IAG cover. Therefore, annual grass cover alone may not be sufficient for
877 quantifying fire risk. Panel a illustrates this point using publicly available data from the
878 Bureau of Land Management's Assessment, Inventory and Monitoring dataset. Panels b and
879 c show quadrats at a site with high, pre-fire native perennial cover weeks before and days
880 after the Hot Pot fire, which burned at high severity at that site.

881 **Figure 3.** Conceptual diagram of the hypotheses tested in this study.

882 **Figure 4.** Panel a is a path model showing support for the various hypotheses depicted in
883 Figure 3. Red arrows are negative relationships, blue arrows are positive relationships, and
884 grey arrows are not significant ($p > 0.05$) but still accounted for in the model. Abbreviations:
885 pre = pre-fire; post = post-fire; cv = cover; elv = elevation; ag = aboveground; sb = seed bank;
886 sev = severity; div = diversity. On the left side of (b), burn severity (dNBR) as predicted by
887 total vegetation cover (TVC; the sum of live and dead, shrub and herbaceous cover). On
888 the right, burn severity is predicted by modelled TVC. In (c), fuel connectivity three years
889 post-fire is modelled by seedbank composition, elevation and pre-fire aboveground species
890 richness. In (d) Shannon-Weaver diversity index of the aboveground, post-fire community

891 composition, was negatively affected by fuel connectivity after accounting for elevation. For
892 a, c and d, lines are the fitted partial effects, points are the partial residuals, and dotted lines
893 are the 95% confidence intervals. $p < 0.05$ for black lines, $p > 0.05$ for grey lines. Panel e
894 shows the modeled occurrence of germinable seeds for all species found at more than one
895 location along a gradient of burn severity, after accounting for soil depth, aspect, elevation
896 and pre-fire diversity. Black line is the mean prediction, each colored line represents one
897 posterior sample.

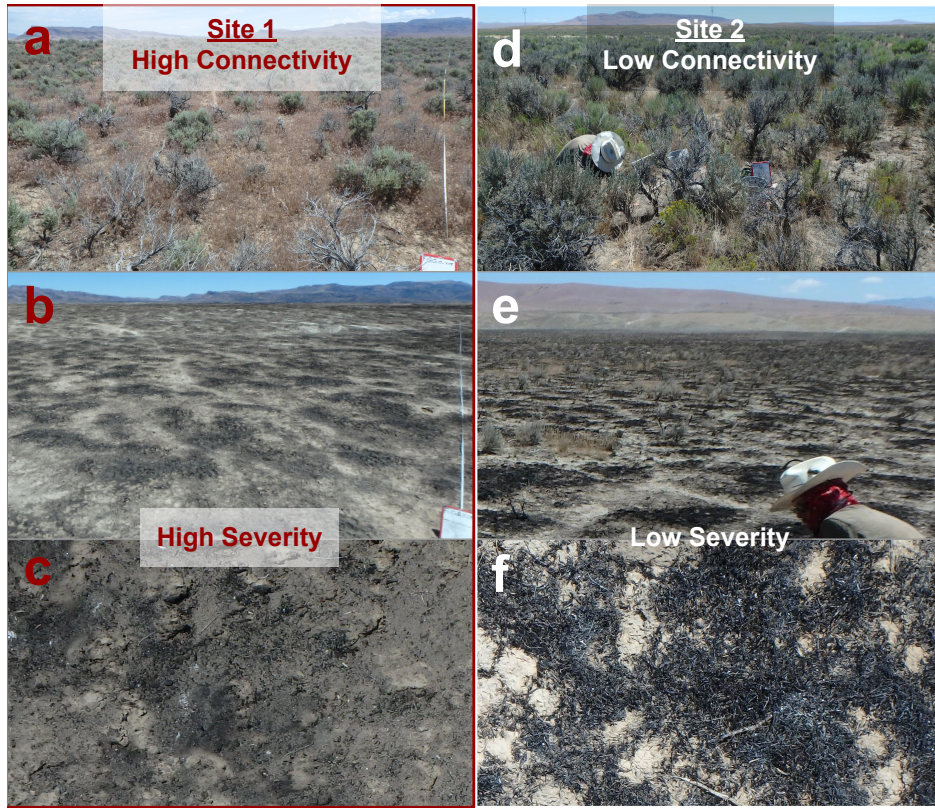


Figure 1: .

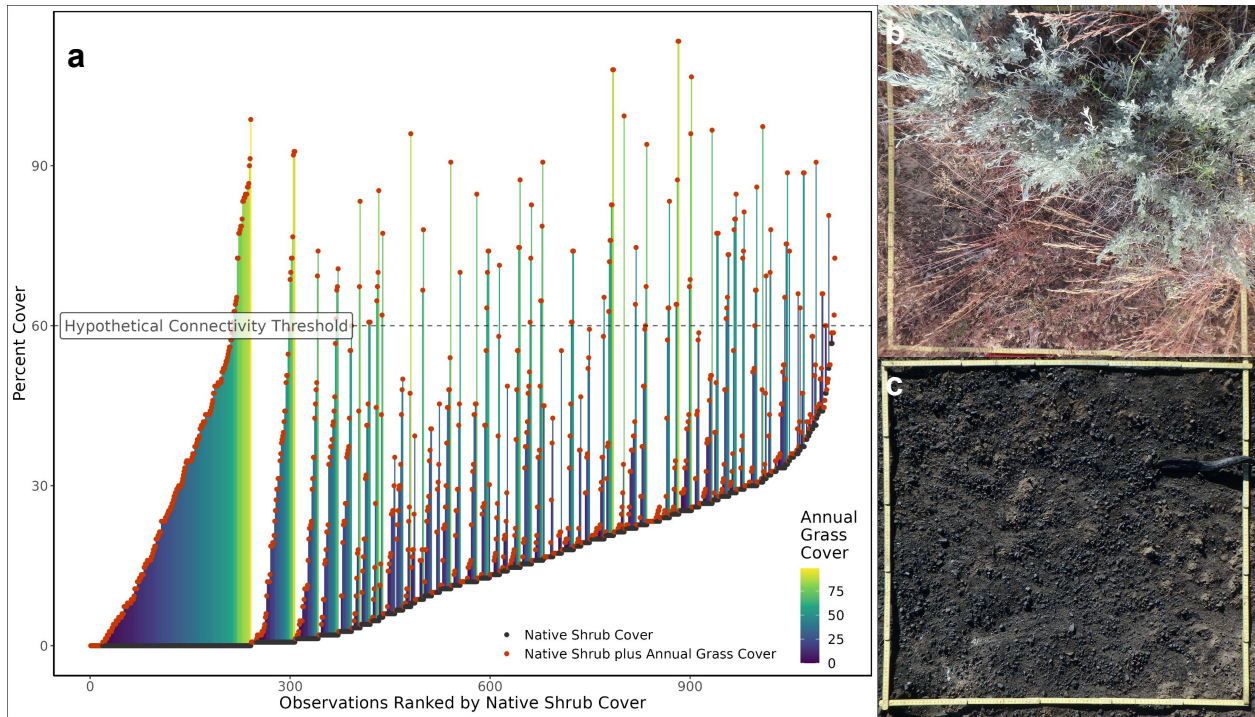


Figure 2: .

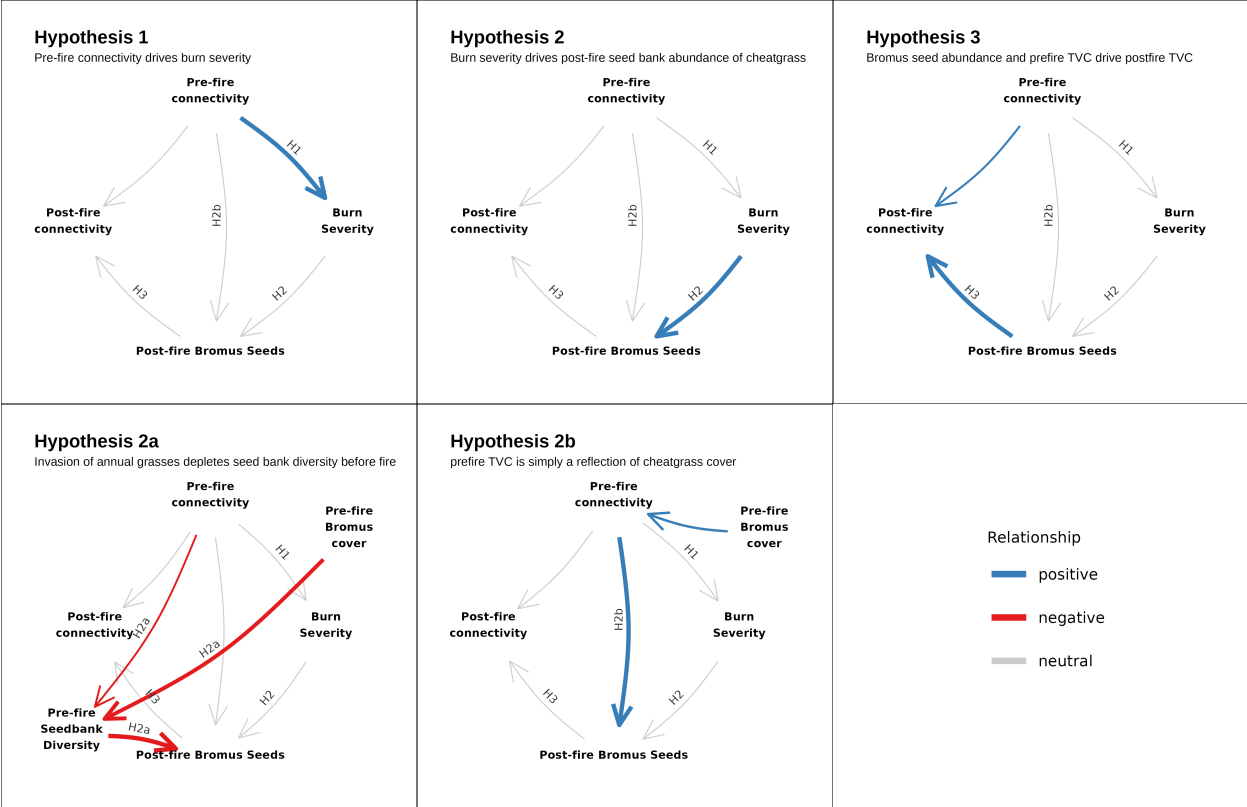


Figure 3: .

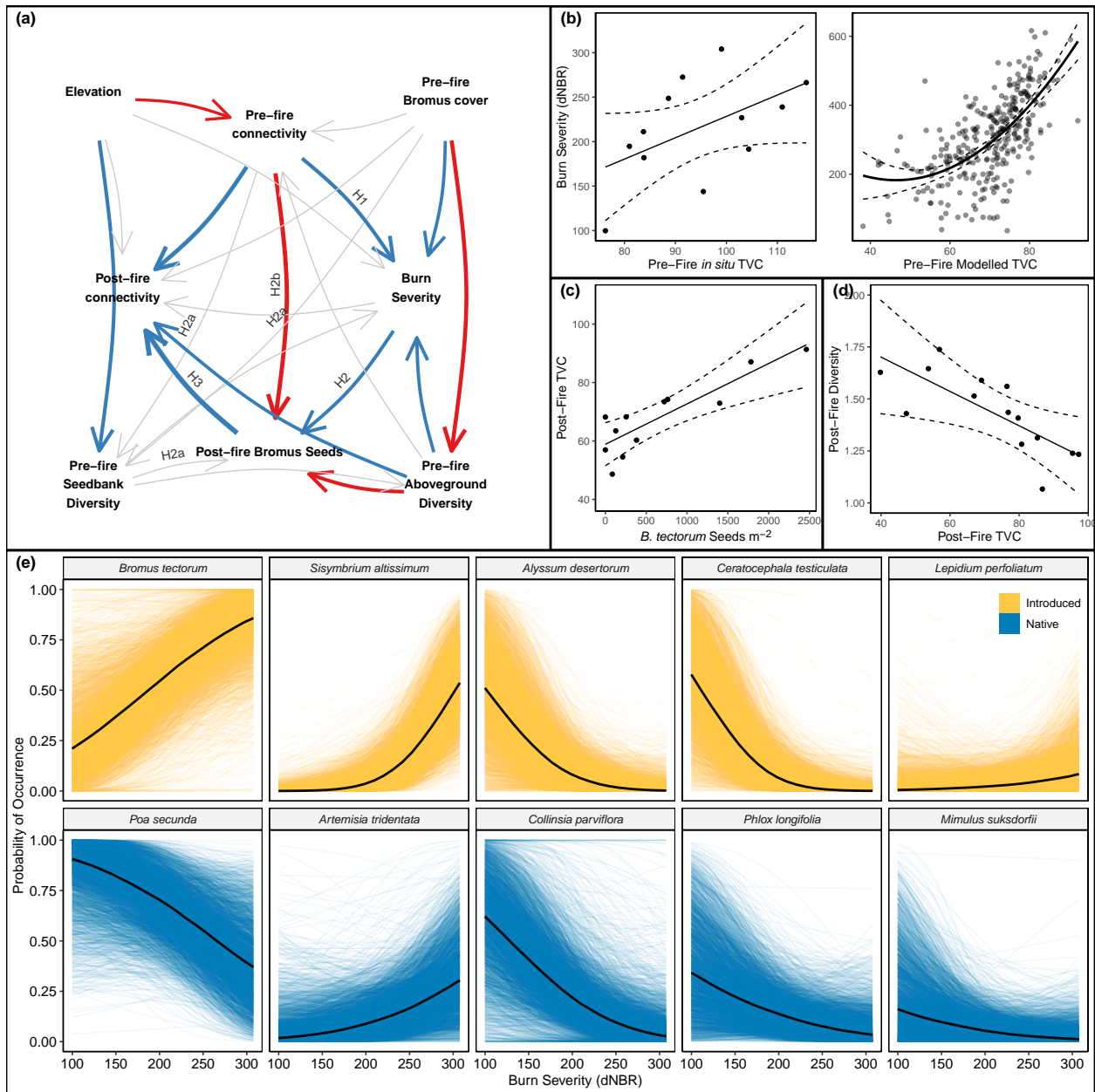


Figure 4: .

Appendix S1 for: “Fuel connectivity, burn severity, and seedbank survivorship drive ecosystem transformation in a semi-arid shrubland.”

Adam L. Mahood^{1,2,3,*}, Michael J. Koontz², Jennifer K. Balch^{1,2,4}

¹ Department of Geography, University of Colorado Boulder, Boulder, CO, USA

² Earth Lab, University of Colorado, Boulder, CO, USA

³ Agricultural Research Service, United States Department of Agriculture, Fort Collins, CO, USA

⁴ Environmental Data Science Innovation and Inclusion Lab, University of Colorado, Boulder, CO, USA

* Corresponding author: admahood@gmail.com

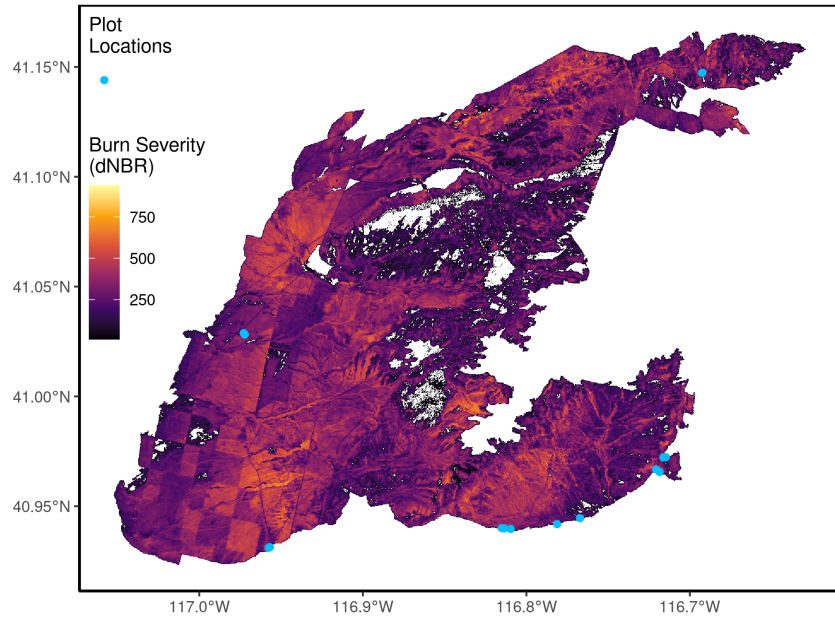


Figure S1: The 2016 Hot Pot Fire. Blue points represent sampling locations and the shaded color is the burn severity. The checkerboard pattern on the lower left corresponds to patterns of land ownership.

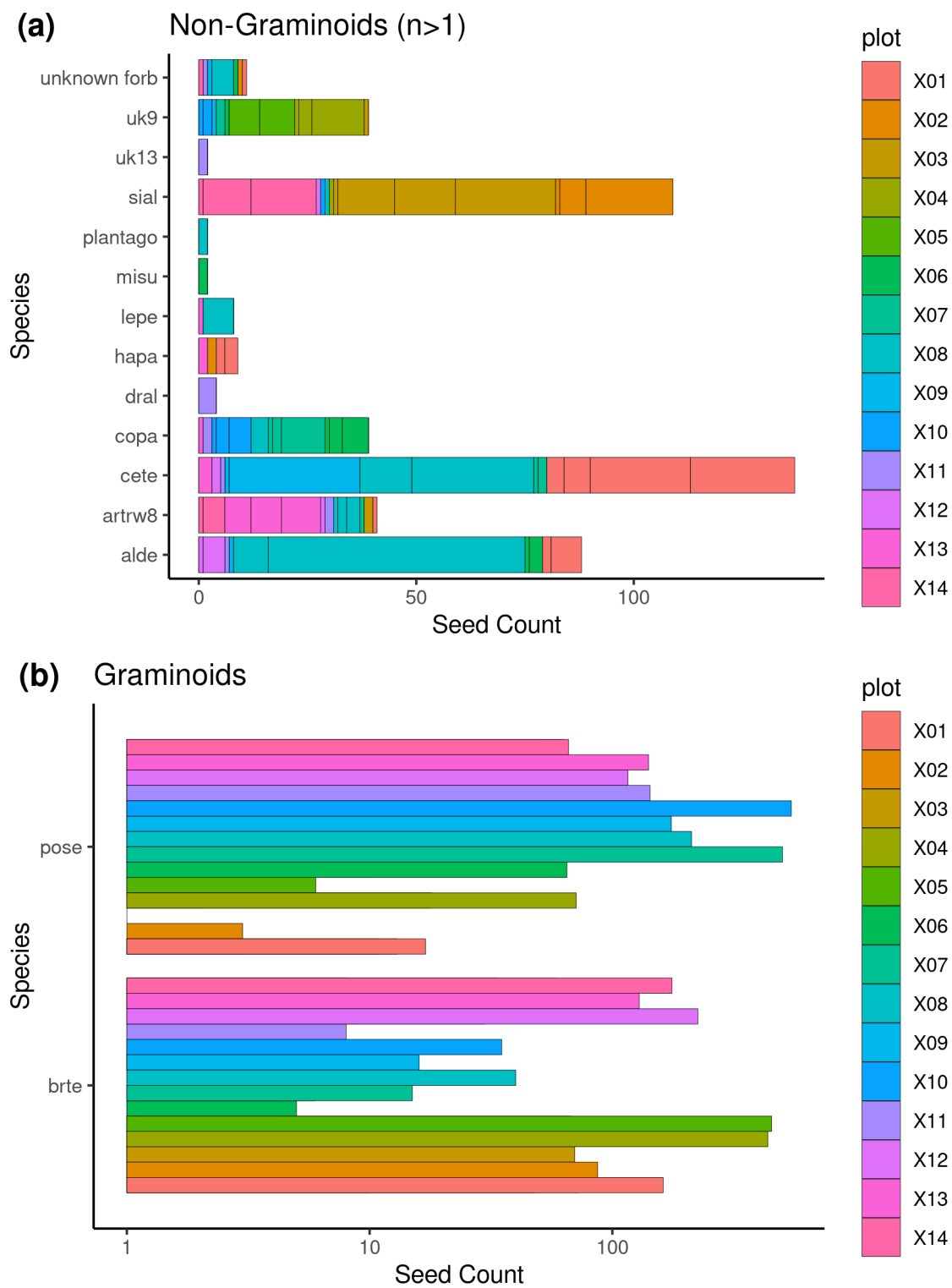


Figure S2: Seed counts by species that occurred more than once. Panel a shows non-graminoids, b shows graminoids.

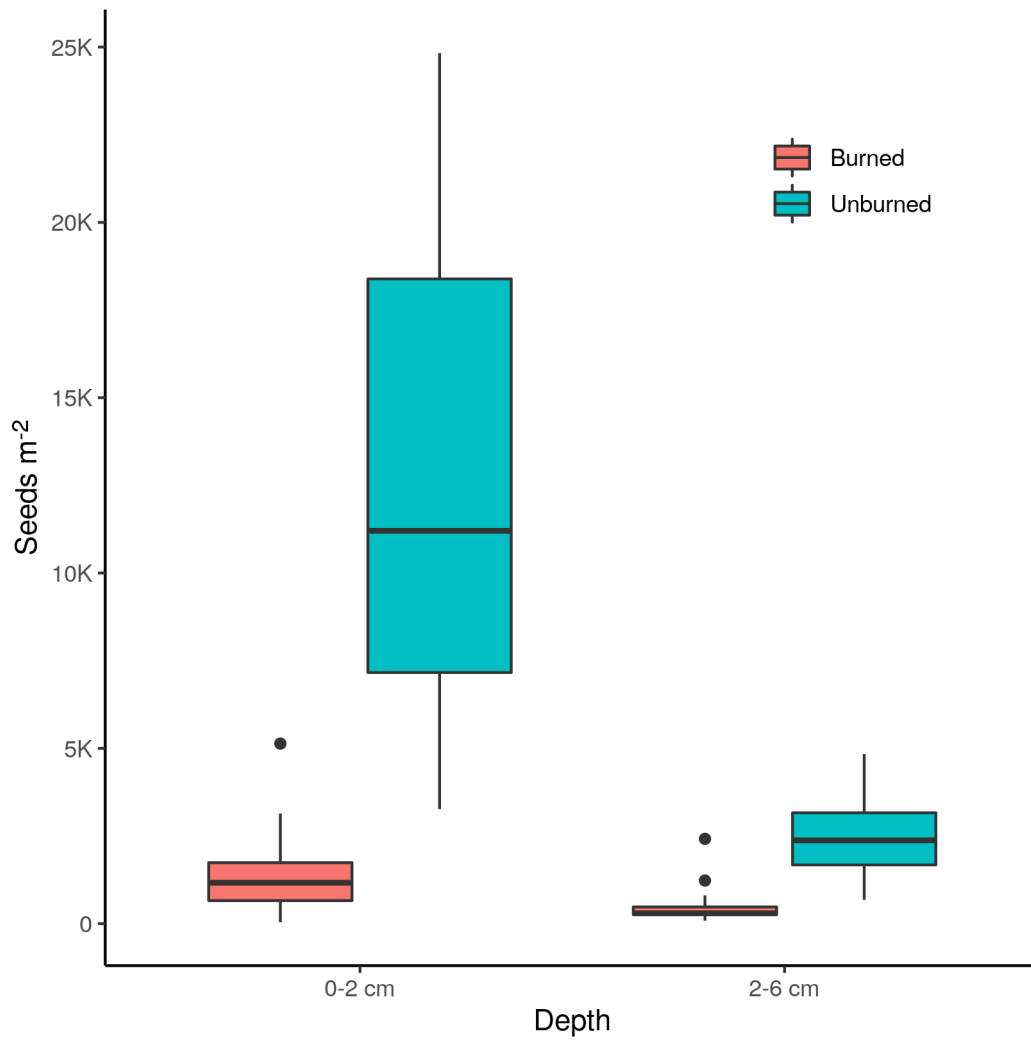


Figure S3: Total seed counts per plot.

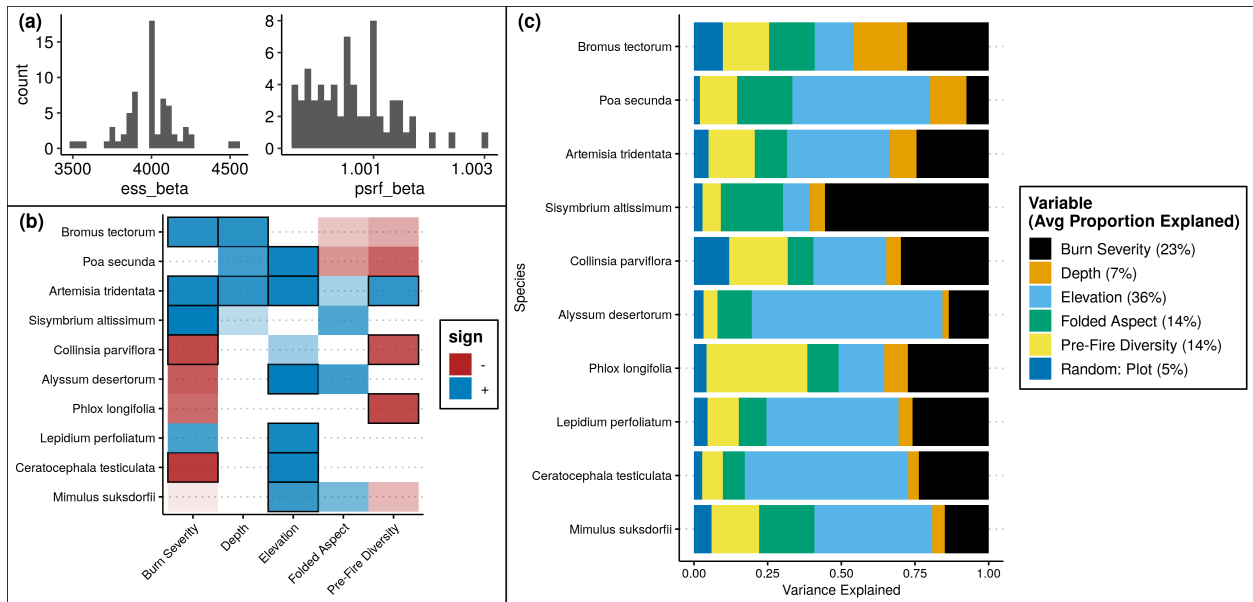


Figure S4: a) Model convergence diagnostics. On the left is the effective sample size after adjusting for autocorrelation (ideally 4,000), and on the right is the Gelman diagnostic, ideally 1. b) Predictor variables that had at least 80% support. Variables with 95% support are outlined in black. The level of transparency corresponds to the level of support. c) Variance partitioning by species. Average across all species per variable is given in the legend. Species are ordered by prevalence.

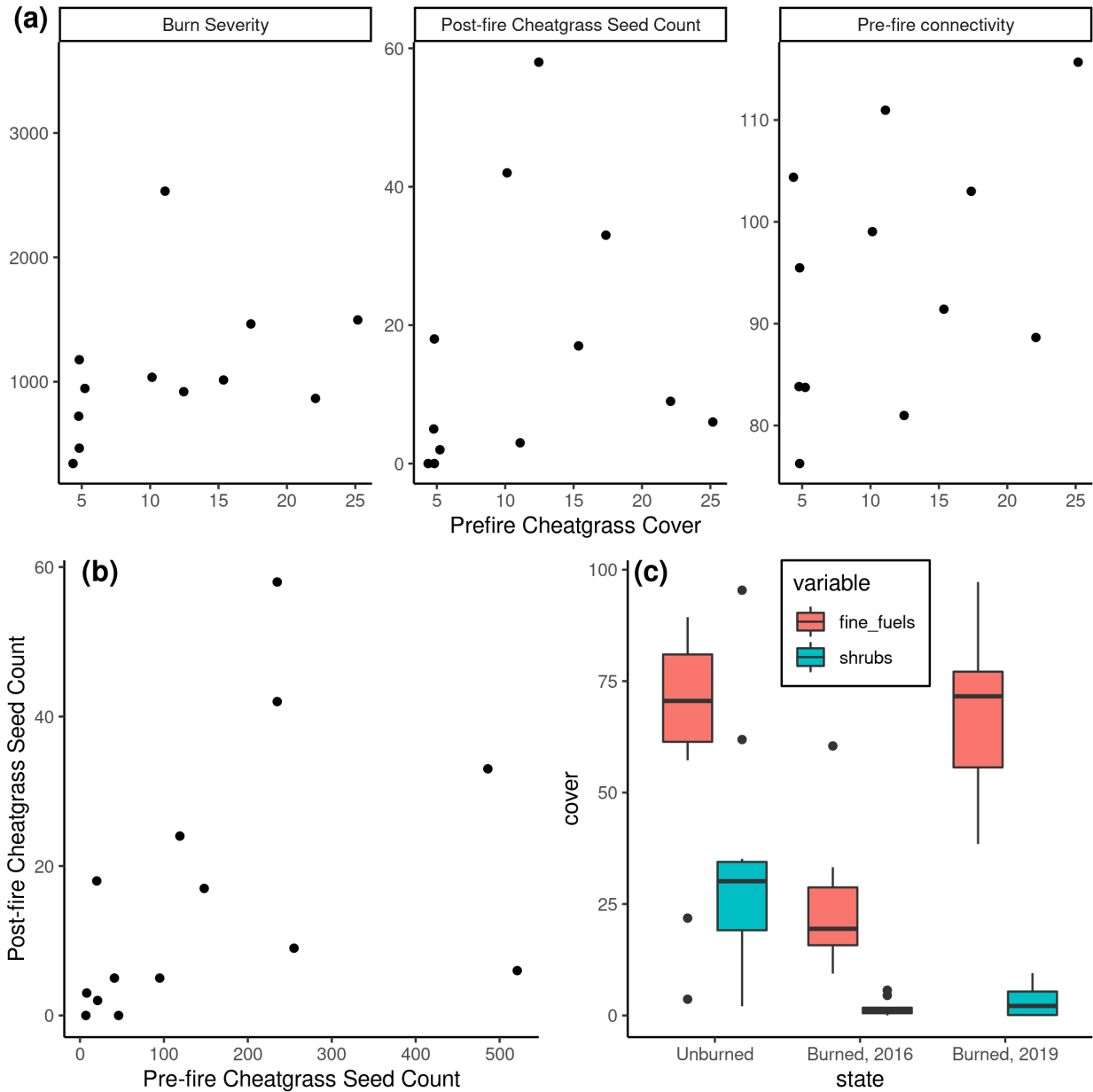


Figure S5: Panel a illustrates how we did not find convincing evidence that pre-fire cheatgrass cover alone was predictive of any of the key components of our hypothesized feedback loop. Panel b shows how even pre-fire cheatgrass seed counts were not predictive of post-fire seed counts. Panel c shows the general change in structural composition, from woody to herbaceous, before and after the fire.

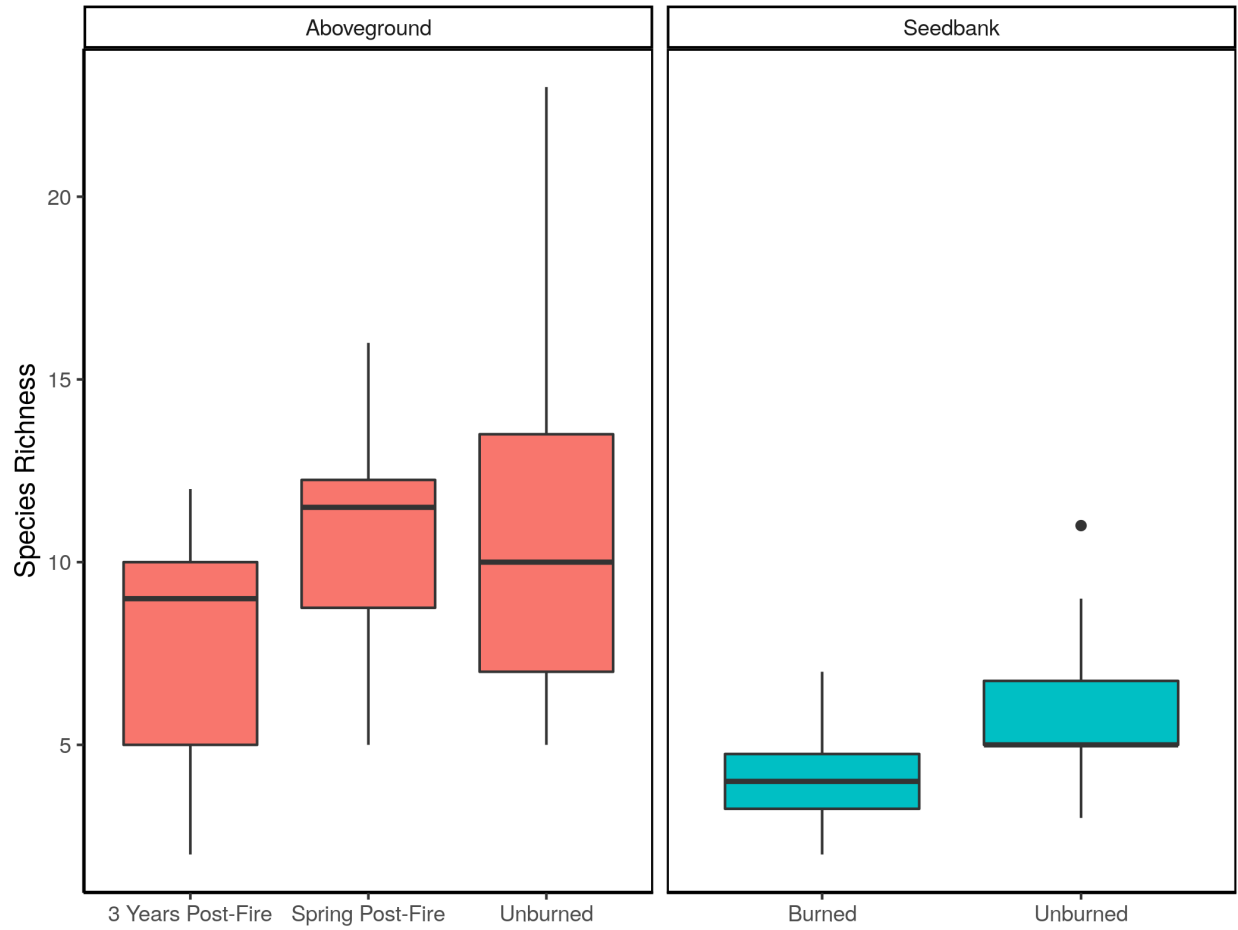


Figure S6: Species richness at different sampling times and locations.

Table S1. Vegetation indexes that were explored in the remote sensing analysis for hypothesis 1.

Index Name	Equation
Green NDVI	$\frac{NIR-Green}{NIR+Green}$
SAVI	$\frac{NIR-Red}{NIR+Red} + 1.5$
NDVI	$\frac{NIR-Red}{NIR+Red}$
EVI	$\frac{NIR-Red}{NIR+(6*Red)-(7.5*Blue)+1} * 2.5$
NDSVI	$\frac{SWIR_1-Red}{SWIR_1+Red}$
NDTI	$\frac{SWIR_1-SWIR_2}{SWIR_1+SWIR_2}$

Table S2: Model performance metrics.

Model	R2	R2_adjusted	Sign
H1: TVC ~ NDSVI + Green NDVI	0.35		+
H1: dNBR ~ TVC(modelled)	0.42	0.42	+
H1: dNBR ~ TVC(in situ)	0.27	0.20	+
H3: Post-Fire Fuel Connectivity ~ # Cheatgrass Seeds + covariates	0.84	0.75	+
H4: Post-Fire Diversity ~ Post-Fire Fuel Connectivity	0.92	0.89	-

Table S3: Seeds germinated in the greenhouse from the cores we collected.

Plot	p1	p2	p3	p4	p5	p6	p7	p8	p9	p10	p11	p12	p13	p14
Burn Severity (dNBR)	195	307	300	226	266	143	211	191	99	181	238	248	272	304
<i>B. tectorum</i>														
U_T2	162	87	70	437	453	5	15	40	16	35	8	225	129	176
U_B4	73	32	25	49	68	2	6	6	4	6	0	30	19	59
B_T2	48	19	4	29	1	0	1	0	15	5	3	9	11	34
B_B4	10	5	1	4	5	0	1	0	3	0	0	0	6	8
<i>P. secunda</i>														
U_T2	17	3	1	71	6	65	502	212	175	546	143	116	141	66
U_B4	13	0	0	18	2	10	55	24	19	49	29	19	29	51
B_T2	11	0	0	2	1	3	21	0	37	32	5	28	8	63
B_B4	3	0	0	0	0	0	4	1	4	4	2	6	18	35
<i>A. tridentata</i>														
U_T2	1	0	0	0	0	0	1	2	0	0	0	1	7	0
U_B4	0	0	0	0	0	0	0	3	0	0	2	0	6	1
B_T2	1	0	2	0	0	0	1	1	0	0	0	0	9	5
B_B4	0	0	0	0	0	0	0	1	0	0	0	0	1	2
<i>A. desertorum</i>														
U_T2	0	0	0	0	0	0	0	59	1	0	0	5	0	0
U_B4	0	0	0	0	0	0	0	8	0	0	1	1	0	0
B_T2	7	0	0	0	0	1	0	0	0	0	0	1	0	0
B_B4	2	0	0	0	0	3	0	0	0	0	0	0	0	0
<i>C. testiculatum</i>														
U_T2	24	0	0	0	0	0	2	28	30	0	1	2	3	0
U_B4	23	0	0	0	0	0	1	12	0	0	0	0	0	0
B_T2	6	0	0	0	0	0	0	0	0	0	0	0	0	0
B_B4	4	0	0	0	0	0	0	0	1	0	0	0	0	0
<i>C. parviflora</i>														
U_T2	0	0	0	0	0	6	10	0	0	3	0	0	1	0
U_B4	0	0	0	0	0	3	0	4	0	1	2	0	0	0
B_T2	0	0	0	0	0	0	2	0	0	3	0	0	0	0
B_B4	0	0	0	0	0	1	1	4	0	5	0	0	0	0
<i>S. altissimum</i>														
U_T2	0	20	23	0	0	0	0	1	0	1	0	0	0	1
U_B4	0	6	13	0	0	0	0	0	0	0	0	1	0	0
B_T2	0	1	14	1	0	0	0	0	0	0	0	0	0	15
B_B4	0	0	1	0	0	0	0	0	0	0	0	1	0	11
<i>M. gracilis</i>														
U_T2	0	0	0	1	0	1	0	0	0	0	0	0	0	0
U_B4	0	0	1	12	8	0	2	0	0	1	0	0	0	0
B_T2	0	0	0	0	0	0	0	0	0	2	0	0	0	0
B_B4	0	0	0	3	7	0	0	1	1	0	0	0	0	0
Other species														
All treatments	9	3	0	0	0	4	0	17	2	0	11	1	11	6

Note:

U = Unburned

B = Burned

T2 = Top 2 cm

B4 = Bottom 4 cm

Table S4: Covariance matrix for the path model.

x	Bromus_seeds_post	prefire_TVC	ag_div_pre	sb_div_pre	burn_sev	postfire_TVC	elv	Bromus_cv_pre
Bromus_seeds_post	0.000	0.006	0.035	0.002	0.019	0.016	-0.075	0.048
prefire_TVC	0.006	0.000	-0.040	-0.003	-0.025	-0.007	-0.005	-0.001
ag_div_pre	0.035	-0.040	0.000	0.000	0.005	-0.012	0.088	0.000
sb_div_pre	0.002	-0.003	0.000	0.000	0.000	0.028	-0.001	0.000
burn_sev	0.019	-0.025	0.005	0.000	0.000	-0.002	0.048	-0.002
postfire_TVC	0.016	-0.007	-0.012	0.028	-0.002	0.000	-0.036	0.046
elv	-0.075	-0.005	0.088	-0.001	0.048	-0.036	0.000	0.000
Bromus_cv_pre	0.048	-0.001	0.000	0.000	-0.002	0.046	0.000	0.000

Table S5: Path model fit measures.

measure	value
degrees of freedom	4.00
p-value	0.92
Chi-Square	0.93
Comparative Fit Index	1.00
Tucker-Lewis Index	1.47
Root Mean Square Error of Approximation	0.00
Standardized Root Mean Square Residual	0.03