

Running head: Burn severity and ecosystem transformation

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

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Open Research Statement: Data and code to recreate the analysis are freely available at <https://www.github.com/admahood/seed-bank> (DOI: <https://doi.org/10.5281/zenodo.5293996>).

1 **Abstract**

2 When multiple drivers interact to affect an ecosystem's function, ecosystem state changes
3 can result. These state changes may be both precipitated and maintained by disturbances,
4 but predicting whether the state change is fleeting or persistent requires an understanding of
5 the mechanisms by which disturbance affects the alternative communities. In the sagebrush
6 shrublands of the western United States, widespread annual grass invasion has increased
7 fuel connectivity, which increases the size and spatial contiguity of fires, leading to post-fire
8 monocultures of introduced annual grasses (IAG). The novel grassland state is persistent,
9 and more likely to promote large fires than the shrubland it replaced. But the mechanisms
10 by which pre-fire invasion and fire occurrence are linked to higher post-fire flammability are
11 not fully understood.

12 We hypothesized that higher fuel connectivity would increase burn severity, which would then
13 have differential effects on post-fire dispersal by causing seed bank survivorship to favor IAG.
14 Subsequent seedbank composition dominated by IAG would lead to increased post-fire fuel
15 connectivity. We found that pre-fire fuel connectivity did increase burn severity. We then
16 used a Bayesian joint species distribution model to examine how burn severity affected the
17 proportion of IAG in the seed bank, and found that higher burn severity had mostly positive
18 effects on the occurrence of IAG and another non-native species, and mostly negative or
19 neutral relationships with all other species. Finally, we found that the abundance of IAG
20 seeds in the seedbank immediately post-fire had a positive effect on the fuel connectivity 3
21 years after fire, completing a positive feedback promoting IAG. These results demonstrate
22 that the strength of the positive feedback is controlled by measurable characteristics of
23 ecosystem structure, composition and disturbance, and each node in the loop is affected
24 independently by multiple global change drivers. It is possible that these characteristics can
25 be modeled to predict threshold behavior and inform management actions to mitigate the
26 undesirable effects of the grass-fire cycle, perhaps via targeted restoration applications or

27 pre-fire fuel treatments.

28 *Keywords:* cheatgrass, sagebrush, grass-fire cycle, joint species distribution model, burn
29 severity, fuel connectivity, *Bromus tectorum*, *Artemisia tridentata*

30 **1. Introduction**

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

49 There is a long history of univariate time series observations that show sudden state changes
50 (Marten Scheffer and Carpenter 2003), and these have informed the development of theories
51 that help us understand how systems of any type can change state suddenly, and exist in

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

70 A central challenge in ecology in the 21st century is to move from describing how plant
71 communities are affected by global change to the capacity to predict how species pools will
72 assemble and persist in response to global change (Davis, Higuera, and Sala 2018; Keddy and
73 Laughlin 2021). Prediction of community response to multi-faceted global change drivers
74 is enhanced with a better understanding of the mechanisms that underlie community sta-
75 bility in the face of disturbances. A classic example of an ecosystem that appears to have
76 disturbance-mediated alternative stable states, but whose stability mechanisms aren’t well
77 understood is the invasion of *Bromus tectorum* L. and other introduced annual grasses in
78 the Great Basin of the western United States. Here, it is well described how the interaction

79 of exotic plant invasions, fire (Balch et al. 2013) and grazing (Williamson et al. 2019) are
80 associated with the degradation or loss of over half of Wyoming big sagebrush (*Artemisia tri-*
81 *dentata* ssp. *wyomingensis* Beetle & Young) ecosystems (Davies et al. 2011). In uninvaded
82 areas, the space between shrubs is typically composed of bare ground covered in biological
83 soil crust and caespitose perennial plants. The lack of fuel connectivity limits fire spread,
84 and the pre-colonial fire regime was one of infrequent, patchy fires (Baker 2006; Bukowski
85 and Baker 2013). Annual grass invasion increases fuel connectivity while decreasing fuel
86 moisture (Davies and Nafus 2013), leading to increased fire size and frequency (Knapp 1996;
87 Balch et al. 2013). After fire, the landscape is typically dominated by introduced annual
88 grasses. But in order to understand how fire drives the persistence of the grassland state, we
89 need to understand the demographic mechanisms by which fire impacts propagule dispersal
90 and benefits the alternative state (Davis, Higuera, and Sala 2018). As with forested systems,
91 propagule dispersal is a key filter through which species must pass in order to establish and
92 persist in a post-fire landscape (Gill et al. 2022).

93 Petraitis and Latham (1999) posited that the maintenance of alternate species assemblages
94 requires first a disturbance that removes the species from the initial assemblage and second
95 the arrival of the species of the alternate assemblage. One understudied mechanism that
96 may explain both for the *Artemisia/Bromus* system is the interaction between the species
97 composition of the soil seed bank and burn severity. Because the invading species are annual,
98 and most of the native plants are seed obligates, the seed is the key life history stage that fire
99 must act upon to benefit the invading plants. Seeds and seedlings are particularly vulnerable
100 to climate, competition and disturbance (Enright et al. 2015). Warmer and drier conditions
101 simultaneously reduce recruitment, growth, and survival of seeds and seedlings (Enright et
102 al. 2015; Schlaepfer, Lauenroth, and Bradford 2014), while also increasing burn severity (S.
103 A. Parks and Abatzoglou 2020). In fire prone ecosystems, seed obligate species typically
104 have life history strategies to cope with fires that burn at different severities (Maia et al.
105 2012; Wright, Latz, and Zuur 2016; Palmer, Denham, and Ooi 2018). Soil heating from fire

106 affects the response of vegetation to fire (Gagnon et al. 2015), including the capacity of
107 seeds to remain viable after fire (Humphrey and Schupp 2001). High severity fire can affect
108 species that use the seedbank positively (Kimura and Tsuyuzaki 2011), negatively (Heydari
109 et al. 2017), or have no effect (Lipoma, Funes, and Díaz 2018), depending on species-specific
110 adaptations. Both the depth of the burn and fire temperature can affect subsequent recovery
111 by seed germination (Morgan and Neuenschwander 1988; Schimmel and Granström 1996),
112 as well as seed mortality and physical seed dormancy mechanisms (Liyanage and Ooi 2017).

113 In addition to size and frequency, exotic plant invasions can alter fire temperature (Brooks et
114 al. 2004; R. O. Jones et al. 2015) and burn severity. While in many cases fires that burn at
115 higher temperatures will also consume more biomass, grass fires may not always have such a
116 relationship. Direct measurements have shown that *B. tectorum* burns at low temperatures
117 (Beckstead et al. 2011; Germino, Chambers, and Brown 2016), but because it also increases
118 horizontal fuel connectivity (Davies and Nafus 2013), it leads to more contiguously burned
119 areas and therefore higher burn severity, despite lower fire temperatures. In order for an
120 annual like *B. tectorum* it to benefit from fire (Balch et al. 2013; Mahood and Balch 2019),
121 it would need enough viable seeds in the post-fire landscape to achieve a fitness benefit
122 and become well-represented in the post-fire plant assemblage (Bond and Midgley 1995).

123 If the fire is patchy, this can happen through post-fire seed dispersal, and *B. tectorum*
124 seeds have been shown to have increased dispersal distances after fire (Monty, Brown, and
125 Johnston 2013). Without unburned patches, seeds must survive the fire. If the increase in
126 fuel connectivity caused by *B. tectorum* increases the severity of fire, it stands to reason that
127 burn severity would then influence the community composition of the post-fire seed bank
128 in a way that facilitates the post-fire dominance of *B. tectorum*. In other words, an area
129 with high burn severity should have a lower occurrence of viable seeds of native species, and
130 a higher occurrence of the seeds of fire-tolerant introduced annual plants. This would lead
131 to the dominance of introduced annual grasses and would result in higher fuel connectivity,
132 closing the positive feedback loop. Plants that are not adapted to frequent fire would be

133 less likely to produce seeds that are adapted to surviving fire, or dispersal mechanisms to
134 take advantage of the resources available immediately after fire (Keeley et al. 2011). To our
135 knowledge, despite several studies on the relationship between fire occurrence and the seed
136 bank in this system (Young and Evans 1975; Hassan and West 1986; Humphrey and Schupp
137 2001; Boudell, Link, and Johansen 2002; Barga and Leger 2018), no studies to date have
138 examined the effect of burn severity on the seed bank.

139 Here, we collected soil cores from 14 locations along the perimeter of a large fire (the Hot
140 Pot fire, ~50,000 ha) immediately after it was extinguished, in northern Nevada in July
141 2016. Because it burned a large area in only three days, we could sample a broad area
142 while being reasonably certain that the climatic conditions during the fire were similar at
143 all sites. Because we collected our samples immediately after the fire was extinguished,
144 we felt confident that the seed bank samples did not contain seeds deposited by post-fire
145 dispersal. We put the samples in cold storage and germinated the seeds from those cores in
146 a greenhouse the following spring. In spring 2017 and fall 2019 we collected information on
147 vegetation structure and diversity at each location. We hypothesized that (H1) Pre-fire fuel
148 connectivity would be positively related to burn severity; (H2) burn severity would increase
149 the occurrence probability of introduced annual species in the seed bank and reduce the
150 occurrence probability of native species. Alternatively, (H2a) increased fuel connectivity
151 brought on by the invasion of annual grasses may have already depleted the diversity of
152 the soil seed bank before the fire occurred; (H3) the abundance of post-fire *B. tectorum*
153 seeds in the seedbank would be positively related to post-fire fuel connectivity. In addition,
154 because in our study system post-fire sites are floristically distinct from the pre-fire state
155 (Mahood and Balch 2019), typically with near monocultures of *B. tectorum*, we hypothesized
156 that (H4) high post-fire fuel connectivity of those near-monocultures would result in lower
157 aboveground species diversity due to competitive exclusion of native plants.

158 2. Methods

159 2.1 Study Area

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

169 2.2 Seed Bank Sampling

170 In early July 2016, we collected samples of the soil seed bank at fourteen locations the day
171 after the Hot Pot fire was contained. Each site was located at the perimeter of the fire where it
172 was clearly delineated by a bulldozer line or in one case a narrow dirt road. Eleven sites were
173 mature sagebrush communities with no history of fire since at least 1984. Three plots had
174 previously burned in 1984 according to the Monitoring Trends in Burn Severity (MTBS) fire
175 history (Eidenshink et al. 2007) and had high cover of *B. tectorum*, but still had scattered
176 sagebrush cover. We used a metal stake to mark paired burned and unburned sampling
177 locations on each side of the perimeter, 10 m from the nearest evidence of anthropogenic
178 disturbance (i.e. bulldozer effects, footprints) associated with active fire suppression along
179 the perimeter. Within 3 m of each marker, we extracted twelve, 6 cm deep, 5 cm diameter,
180 soil cores. Seeds of sagebrush generally do not fall far (<30 m) from their parent plants in this
181 system (Shinneman and McIlroy 2016), and so they are not uniformly distributed (Boudell,
182 Link, and Johansen 2002). In addition, seeds from *B. tectorum* (Young and Evans 1975)
183 and *Artemisia* have different germination rates based on the micro-site they find themselves

184 in (i.e. under a shrub or in the bare ground between shrubs, [Eckert et al. 1986](#)). To account
185 for these potentially confounding effects, we placed half of the core locations under shrubs,
186 and half in shrub interspaces. In the burned areas, it was obvious where shrubs had been
187 located. Even when they were completely incinerated, their imprint remained on the soil
188 surface ([Bechtold and Inouye 2007](#); [Germino et al. 2018](#)). To examine the effect of seed
189 depth, we divided each soil core into 0-2 cm and 2-6 cm depths. Litter was aggregated with
190 the 0-2 cm samples. Samples were then placed in cold storage (~2 deg C) for 3 months
191 ([Meyer, Monsen, and Mcarthur 2013](#)). At all sites, to be sure that we were at a site where
192 sagebrush germination could occur we checked for first year germinants on the unburned side
193 (we found them at all sites), and to ensure that there were no confounding effects of post-fire
194 seed dispersal, we determined whether or not the sagebrush were flowering (they were not
195 flowering at all sites), and recorded species occupancy for all aboveground plant species.

196 We followed the methodology of Ter Heert et al. ([1996](#)) to germinate the seeds. Each
197 sample was run through 0.2 mm sieve, and spread in a 3-5 mm layer over the top of 1 - 4
198 pots. These pots were filled 3 cm deep with potting soil, topped by a thin layer of sand.
199 Pots were watered as needed to stay at field capacity. Every week emerging germinants were
200 identified, counted and removed. Most of the germination occurred within 6 weeks, and after
201 8 weeks we ended the germination assay.

202 *2.3 Post-Fire Vegetation Sampling*

203 We sampled the aboveground fuel structure and plant diversity in May 2017, the growing
204 season immediately after the fire and again in September 2019. At each location, we es-
205 tablished 50m transects starting at the boundary of the burned and unburned sides of the
206 perimeter, running perpendicular to the fire perimeter, and marked the transect ends with
207 rebar. In order to characterize aboveground plant diversity, we measured the occupancy and
208 abundance of all plant species by measuring cover of every species in 0.1 m² quadrats spaced
209 every 5 m along each transect. In order to characterize fuel structure, we used the line in-

210 tercept method to measure shrub cover (coarse fuels) and herbaceous plant cover (fine fuels)
211 along the transect, and summed those measurements to get total vegetation cover (TVC).
212 Both live and dead plants were included in these measurements.

213 *2.4 Estimating Burn Severity with Landsat 8 OLI*

214 We downloaded the “fire bundle” of the Hot Pot fire from www.mtbs.gov. This included
215 cloud-free Landsat 8 scenes collected before the Hot Pot fire, and already calculated layers
216 of the Differenced Normalized Burn Ratio (dNBR, [J. D. Miller et al. 2009](#)). Because our sites
217 were generally within 10 meters of the burn perimeter, The pixels directly intersecting the
218 plot locations were likely to be mixed pixels (i.e. containing burned and unburned ground).
219 To minimize this effect, we extracted all the dNBR values within a 120 meter buffer of each
220 seed bank plot for pixels whose centroids fell inside of the fire perimeter and calculated the
221 mean.

222 *2.5 Statistical Analysis*

223 Our statistical analysis centered around trying to understand each component of the positive
224 feedback loop posited by the 4 hypotheses described above. In order to understand how pre-
225 fire fuel connectivity influenced burn severity (H1), we used total vegetation cover (TVC)
226 from two separate data sources as a proxy for fuel connectivity, and created separate linear
227 models with TVC as the predictor variable and burn severity (dNBR, [J. D. Miller et al.](#)
228 [2009](#)) as the response variable. With the field data we collected, we created an ordinary
229 least squares (OLS) linear model with burn severity as the dependent variable and TVC
230 (defined as shrub cover plus herbaceous plant cover from the unburned side of the paired
231 plots), elevation and aspect as independent variables.

232 We were concerned that because our data were collected at the edge of the fire, the burn
233 severity calculated at each point may have included partially burned pixels. So, as a sup-
234 plement, we examined the same relationship by creating a model of TVC using Landsat
235 Thematic Mapper (TM) surface reflectance data using TVC from the Bureau of Land Man-

236 agement’s Assessment, Inventory and Monitoring dataset (AIM, [U.S. Department of Inte-](#)
237 [rior 2018](#)). The AIM dataset contained 813 sampling locations within the Central Basin
238 and Range ecoregion ([Commission for Environmental Cooperation 2006](#)) that were visited
239 by BLM field crews between 2011 and 2015. They were mostly sampled once but there
240 were some repeats, for 1,117 total measurements. For each of these points, we extracted
241 the surface reflectance values of each Landsat band for the sampling year near peak biomass
242 using a cloud-free scene from May or early June. Then, we used those surface reflectance
243 values to calculate various vegetation indexes (Appendix S1: Table S1), including the Green
244 Normalized Differenced Vegetation Index (Green NDVI, Equation 1), and Normalized Dif-
245 ferenced Senesced Vegetation Index (NDSVI, Equation 2). We used these indexes to create
246 generalized linear model of TVC with a beta distribution. For this and all linear models, we
247 started with the largest possible model and used backwards selection following the method-
248 ology of [Zuur et al. \(2009\)](#). We used the final reduced model to create a layer of predicted
249 TVC for the study area for the pre-fire scene, and extracted both our predictions of TVC
250 and dNBR of the fire from 1000 regularly-spaced points within the fire perimeter. Finally,
251 to quantify the effect of TVC on burn severity, we created an OLS linear model with our
252 modeled TVC and its second-order polynomial as predictor variables and burn severity as
253 the response variable.

254 **Equation 1:** $Green\ NDVI = \frac{NIR-Green}{NIR+Green}$

255 **Equation 2:** $NDSVI = \frac{SWIR_1-Red}{SWIR_1+Red}$

256 To examine how burn severity affected the community composition of the seed bank (H2),
257 we created a joint species distribution model (JSDM) in a Bayesian framework ([Tikhonov](#)
258 [et al. 2020](#)) for the occurrence of all species germinated from the seed bank that were
259 found at more than one location. We created four Markov Chain Monte Carlo (MCMC)
260 chains, each consisting of 150,000 iterations. We discarded the first 50,000 iterations for
261 each chain and then recorded every 100th for a total of 1,000 posterior samples per chain,

262 and 4,000 total. We assessed model convergence using the effective sample size and the
263 potential scale reduction factor (Gelman, Rubin, et al. 1992). We used the model to predict
264 the probability of occurrence of germinable seeds of a given species along a gradient of burn
265 severity. We included burn severity, elevation, aspect, pre-fire seedbank diversity and soil
266 depth as independent variables.

267 To account for the possibility of H2a as a confounding factor, we included the Shannon-
268 Weaver diversity index (Shannon and Weaver 1949) in the paired, unburned seed bank
269 samples as one of the predictor variables in our JSJM. We also created OLS models with
270 the unburned species richness and Shannon-Weaver diversity index predicted by prefire fuel
271 connectivity, with the expectation that pre-fire fuel connectivity would have had a negative
272 effect on the prefire seedbank diversity. To examine how community composition and burn
273 severity then affected subsequent fuel connectivity (H3), we created OLS models with fuel
274 connectivity three years post-fire as the dependent variable, and burn severity, seed counts
275 for *B. tectorum*, *P. secunda* and other species, elevation, aspect, depth, and alpha diversity
276 as independent variables. To examine how the resulting fuel connectivity was related to
277 biodiversity (H4), we used the diversity data and connectivity data that we collected in
278 2019 to create a Poisson GLM with number of species encountered aboveground at each plot
279 location as the dependent variable, as well as an OLS linear model with the Shannon-Weaver
280 index (Shannon and Weaver 1949) as a dependent variable. We used fuel connectivity,
281 elevation, and aspect as independent variables.

282 All analyses were done in R (R Core Team 2020). Data and code to recreate the analysis
283 are freely available at <https://www.github.com/admahood/seed-bank> (DOI available after
284 acceptance) and the Dryad data repository (url given after acceptance).

285 3. Results

286 We found support for each hypothesized component of the positive feedback loop. For H1,
287 the most parsimonious model for our *in situ* observations had only TVC as the predictor,
288 which had a weak positive relationship with burn severity ($\beta = 2.4$, $p = 0.083$, $R^2 = 0.27$,
289 Figure 1a, Appendix 1: Table S2). For our remotely sensed analysis, our most parsimonious
290 model of TVC explained 35% of the variation and had Green NDVI, NDSVI and their
291 interaction as predictors (Appendix S1: Table S2). Our model of dNBR using the predicted
292 TVC within the fire perimeter as a predictor of dNBR explained 42% of the variation and the
293 relationship between TVC and burn severity was positive and significant ($p \ll 0.01$, Figure
294 1a, Appendix S1: Table S2).

295 The vast majority of seeds that germinated in the greenhouse were the two most common
296 grass species, *P. secunda* and *B. tectorum* (Appendix S1: Table S3). Eight dicot species were
297 found in more than one location, and these 10 prevalent species are those that were used
298 in our JSDM. Burned plots had an average of 34 ± 32 total seeds in the top 2 cm, and 12
299 ± 14 in the bottom 4 cm. Unburned plots had an average of 299 ± 170 in the top 2 cm
300 and 59 ± 29 in the bottom 4 cm (Appendix S1: Fig. S2). For H2, the JSDM converged
301 well (Appendix S1: Fig S3). Gelman diagnostics were all very close to 1 and the effective
302 sample size centered on 4,000, which indicated good model convergence. Elevation had the
303 strongest effects on individual species occurrence and explained the most variance on average
304 (36%). Burn severity explained 23% of the variance on average and was supported at the
305 95% level for 5 species. For the introduced species, the predictions along a gradient of burn
306 severity were positive for *B. tectorum*, *Sisymbrium altissimum* L. and *Lepidium perfoliatum*
307 L., and negative for *Ceratocephala testiculata* and *Alyssum desertorum* Stapf. For native
308 species, the effect of burn severity on occurrence was positive for *A. tridentata*, but the
309 mean prediction never rose above 50%. It was neutral for *P. secunda* and negative for the
310 remaining species. Testing H2a revealed a positive relationship between pre-fire aboveground

311 species diversity and pre-fire fuel connectivity, and so we felt it was reasonable to rule out
312 pre-fire fuel connectivity as a confounding factor for H2.

313 For H3, we found that, after accounting for elevation, pre-fire aboveground richness, and
314 the number of *P. secunda* seeds, the number of *B. tectorum* seeds in the post-fire seedbank
315 was positively associated with the fuel connectivity in 2019 ($\beta = 0.54$, $p = 0.01$, Adj R^2
316 $= 0.75$, Figure 1c, Appendix S1: Table S2). For H4 the most parsimonious model (Adj R^2
317 $= 0.89$, Appendix S1: Table S2) had elevation, aspect, fuel connectivity and an interaction
318 between elevation and fuel connectivity as predictors of aboveground Shannon-Weaver alpha
319 diversity. Fuel connectivity was negatively associated with Shannon-Weaver diversity ($\beta =$
320 -0.28 , $p=0.004$, Figure 1d).

321 4. Discussion

322 Here we document the mechanisms by which changes in ecosystem structure brought on by
323 invasion can lead to cascading effects on ecosystem function and composition via changes
324 in the disturbance regime. It has already been shown that *B. tectorum* invasion increases
325 fire frequency (Balch et al. 2013), and is indicative of a grass-fire cycle. However, an
326 understanding of the positive feedback mechanisms that link *B. tectorum* invasion success to
327 fire occurrence is required to infer the long-term persistence of such a cycle. The interaction
328 between burn severity and seed bank composition documented here may explain that link.
329 Prior work has shown that annual grass invasion increases fuel connectivity by filling in
330 shrub interspaces with a contiguous bed of fine fuels (Davies and Nafus 2013). This change
331 in the spatial distribution of fine fuels has been associated with larger and more frequent
332 fires (Balch et al. 2013). Here, we found higher fuel connectivity (via TVC) increased burn
333 severity (H1, Figure 1a). Higher burn severity was associated with an increased occurrence of
334 introduced annuals in the seedbank and a decreased occurrence of native plants (H2, Figure
335 1b). Finally, greater abundance of *B. tectorum* seeds in the post-fire seedbank resulted

336 in higher post-fire fuel connectivity (H3, Figure 1c). In addition, we found evidence that
337 high post-fire fuel connectivity was associated with lower aboveground diversity (H4, Figure
338 1d). This suggests that during inter-fire intervals, there may be additional mechanisms
339 (e.g. competition, altered ecohydrology) maintaining the post-fire, annual grass-dominated
340 species assemblage.

341 The difference in species composition before and after fire explains an apparent contradiction
342 in results between H2a (positive relationship between pre-fire fuel connectivity and diversity)
343 and H4 (negative relationship between post-fire fuel connectivity and diversity). Most plot
344 locations had mature canopies of native shrubs with the inter-shrub space occupied mostly
345 by native bunchgrasses and forbs, with no fire occurrence since 1984. Even in locations with
346 high annual grass cover between shrubs, shrubs provide ecosystem structural heterogeneity
347 and islands of fertility (Doescher, Miller, and Winward 1984; Bechtold and Inouye 2007), and
348 perennial natives that may have been established before invasion have deep roots established
349 that allow for the avoidance of competition for water with shallow-rooted annuals (Gibbens
350 and Lenz 2001; Ottaviani et al. 2020). This may provide enough niche compartmentalization
351 to allow native plants to persist in spite of the invasion. Three years after fire, almost all of
352 the plots were dominated by introduced annuals, and lacked any structural heterogeneity.
353 Thus native plants may have been able to persist via niche compartmentalization after
354 the initial invasion, but fire burned away most of the seeds (Appendix S1, Fig. S1) and
355 removed all of the structural benefits, and microclimatic refugia that shrub cover provides.
356 In this clean slate post-fire environment, the altered species composition of the seedbank
357 and superior post-fire dispersal of *B. tectorum* (Monty, Brown, and Johnston 2013) allow
358 the process of interspecific competition to be dominant (Schlaepfer, Lauenroth, and Bradford
359 2014).

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

361 Burn severity metrics like dNBR were conceived of in the context of forested ecosystems,

362 and calibrated using the composite burn index (Key and Benson 1999), tree mortality, and
363 percent change in tree canopy cover (J. D. Miller et al. 2009). It is unclear how well
364 these metrics carry over to shrubland systems. We recorded qualitative observations of burn
365 severity while we were sampling, mainly to ensure that we sampled a range of severities,
366 and the dNBR we used appears to correspond with our observations. In areas where the
367 space between shrubs was well-connected by fine fuels (Figure 2 a-c) the dNBR was higher,
368 and the shrubs had completely burned throughout the root system, leaving only a hole in
369 the ground filled with ashes as evidence of their prior presence. In these areas the entirety
370 of the soil surface—underneath shrub canopy and in canopy interspaces—was consumed by
371 fire, and there was little evidence of remaining litter or biological soil crust. Areas with
372 lower fuel connectivity had lower dNBR (Figure 2 d-f). Here, shrubs were usually consumed
373 only to the stumps, and sometimes left standing and charred, destined for mortality. In
374 these areas the soil surface often still had biological soil crust, partially consumed litter
375 (R. O. Jones et al. 2015) and unconsumed annual and perennial grass bases. The manual
376 severity classification provided by MTBS had exclusively low and medium severity, but our
377 observations of essentially complete consumption of plant and litter tissues and very few
378 unburned patches suggested that these should have been mostly medium and high severity.
379 This discrepancy was not unexpected, as the ordinal burn severity classifications produced
380 by MTBS are known to be of limited use for research (Kolden, Smith, and Abatzoglou 2015).
381 Spectral reflectance has long been used to characterize ecosystem structure, including wildfire
382 fuels. Unique signatures of remotely-sensed spectral reflectance are typically matched to
383 categorical fuel classifications (CFCs), which describe the physiognomy of vegetation and
384 its potential to support various fire behavior (Ottmar et al. 2007). While different CFCs
385 can provide a general understanding of fuel amount and connectivity, recent efforts using
386 data with finer spatial and spectral resolution may improve fuel classification with more
387 continuous, multi-dimensional measurements (Stavros et al. 2018). The continuous measure
388 of NDVI in western U.S. coniferous forests is a proxy for live fuel biomass, which likely

389 explains its positive association with wildfire severity (Sean A. Parks et al. 2018; Koontz
390 et al. 2020). NDVI also correlates with vegetation cover in these forested systems, and so
391 greater crown connectivity may also explain the NDVI/severity relationship at local scales.
392 When using a more direct NDVI-derived measure of vegetation connectivity in Sierra Nevada
393 yellow pine/mixed-conifer, Koontz et al. (2020) found that greater variability in forest
394 structure, implying fuel discontinuity decreased the probability of high-severity fire. Here,
395 we arrived at a combination of NDVI and NDSVI to describe the fuel connectivity of the
396 annual grass invaded Great Basin sagebrush community to better reflect key differences in
397 the physiognomies of forest and arid shrublands. In sagebrush shrublands, the fuel that
398 contributes to large wildfires is a mixture of evergreen shrubs interspersed with herbaceous
399 plants that remain green for only a portion of the growing season, and then become dry and
400 straw-colored. Thus, both the live and dead fuel need to be taken into account in remote
401 measurements of fuel connectivity for this system.

402 *Management implications*

403 These results demonstrate that the strength of the grass-fire cycle in this system is controlled
404 by measurable fire properties and ecosystem structural components. Land managers may
405 be able to increase their chances of restoration success by using existing methods or devel-
406 oping novel ones that manipulate these components to weaken or even break the positive
407 feedback cycle. This work provides further evidence that the post-fire annual grassland is
408 a system where the degraded state represents an alternative species assemblage from that
409 of the restoration target. Because the propagules of the original assemblage are no longer
410 present, methods that rely on natural succession may not be sufficient (Suding, Gross, and
411 Houseman 2004). Estimating burn severity using satellite imagery may help land managers
412 identify areas with a greater likelihood of successful seeding. Our results highlight the impor-
413 tance of prioritizing the preservation of existing native shrub cover and in particular policies
414 that encourage land managers to maximize the preservation of unburned patches within the
415 fire perimeter during the suppression of wildfires in this system (Steenvoorden et al. 2019),

416 as these are the primary sources of native propagules.
417 Livestock grazing can reduce fuel connectivity in uninvaded sagebrush (Davies et al. 2010).
418 At the same time, livestock grazing can decrease the resistance to invasion by *B. tectorum* via
419 negative effects on biological soil crust (BSC) (Chambers et al. 2014; Condon and Pyke 2018),
420 and can reduce the survival of *Artemisia* seedlings that are not protected by shrub canopies
421 (Owens and Norton 1992). In wet years, targeted grazing at already invaded sites may
422 reduce fuel connectivity and alleviate fire risk. Post-fire grazing may help reduce *B. tectorum*
423 cover, but it may also exacerbate the problem by introducing cheatgrass in uninvaded sites
424 (Williamson et al. 2019) or increasing the already superior post-fire dispersal of *B. tectorum*
425 seeds (Monty, Brown, and Johnston 2013). We suggest management approaches that are
426 specifically tailored each year to the conditions of a given site, and targeting grazing only
427 in already invaded areas to reduce *B. tectorum* cover where it may aid in native plant
428 restoration.

429 Herbaceous cover in these dryland systems has high interannual variability (Mahood et al.
430 2022). Because the components of ecosystem structure and disturbance severity in posi-
431 tive feedback cycle described here are continuous mechanistic variables, it may be possible
432 to develop theoretical models (*sensu* (Ludwig, Jones, and Holling 1978)) to estimate the
433 threshold of vegetation cover that will lead to high burn severity. These can then be applied
434 in conjunction with near real time fuel loading forecasts (M. O. Jones et al. 2021) to identify
435 areas that are vulnerable to high severity fire, which can be used by land managers to take
436 preemptive measures in high value areas.

437 ***Global implications***

438 Understanding how different facets of global change create multiple mechanisms that act in
439 concert to drive ecosystem transformation will provide important insights about ecosystem
440 change on a global scale. Ecological systems are much more complex than a simple bivariate
441 system with a single driver and a single response. The system studied here has at least

442 four external processes that may influence the positive feedback loop we documented. First,
443 land use change in the form of the introduction of livestock grazing facilitates invasion
444 (Ponzetti, Mccune, and Pyke 2007; Williamson et al. 2019). Second, the introduction of
445 exotic grasses increases fuel connectivity (Davies and Nafus 2013), which we document affects
446 burn severity. Third, increasing temperatures due to climate change both increase burn
447 severity (S. A. Parks and Abatzoglou 2020) while simultaneously decreasing seed viability and
448 seedling survival (Schlaepfer, Lauenroth, and Bradford 2014; Enright et al. 2015). Fourth,
449 CO₂ enrichment may preferentially enhance biomass (i.e. higher fuel connectivity) and seed
450 production of annual grass species (Smith et al. 2000; Nagel et al. 2004), strengthening
451 the fuel connectivity to burn severity to seed composition feedback loop. All four of these
452 external drivers are globally ubiquitous consequences of global change.

453 The “state” the ecosystem is the product of countless endogenous interactions. The grass-
454 fire cycle studied here is reinforced through providing fitness benefits to the introduced
455 annual grasses via at least three redundant processes. First, we document how it changes
456 the composition of the seedbank. Second, introduced annual grasses competitively exclude
457 native plants. Third, the dominance of introduced annual grasses initiates ecohydrological
458 feedbacks to create a warmer, drier microclimate (Turnbull et al. 2012; Wilcox et al. 2012).
459 It is possible that some of these feedbacks are idiosyncratic to the system being studied,
460 while others may reflect fundamental properties of ecosystem function that change when a
461 system is converted from being dominated by deep-rooted woody plants to being dominated
462 by annual herbaceous plants (Kitzberger et al. 2016). At least 13 grass species initiate self-
463 reinforcing feedbacks with fire in the U.S. alone (Fusco et al. 2019; Tortorelli, Krawchuk, and
464 Kerns 2020). There are likely many more worldwide, beyond documented cases in Australia
465 (G. Miller et al. 2010; Setterfield et al. 2010), Brazil (Rossi et al. 2014) and South Africa
466 (Milton 2004). The conversion of forests and shrublands to grasslands may have consequences
467 relevant to the global carbon cycle, especially when ecosystems dominated by deep-rooted
468 plants that store carbon belowground are replaced by shallow-rooted ecosystems that lose

469 carbon to grazers and fire [Mahood et al. \(2022\)](#).

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812 **Figure Captions**

813 **Figure 1.** On the left side of (a), burn severity (dNBR) as predicted by total vegetation cover
814 (TVC; the sum of live and dead, shrub and herbaceous cover). On the right, burn severity is
815 predicted by modelled TVC. Panel b shows the modelled occurrence of germinable seeds for
816 all species found at more than one location along a gradient of burn severity, after accounting
817 for soil depth, aspect, elevation and pre-fire diversity. Black line is the mean prediction, each
818 colored line represents one posterior sample. In (c), fuel connectivity three years post-fire is
819 modelled by seedbank composition, elevation and pre-fire aboveground species richness. In
820 (d) shannon-Weaver diversity index of the aboveground, post-fire community composition,
821 was negatively affected by fuel connectivity after accounting for elevation. For a, c and d,
822 lines are the fitted partial effects, points are the partial residuals, and dotted lines are the
823 95% confidence intervals. $p < 0.05$ for black lines, $p > 0.05$ for grey lines.

824 **Figure 2.** Visual illustration of the relationship between fuel connectivity and burn severity.
825 On the left, panel a shows the intershrub space invaded by annual grasses. The photo in
826 panel b was taken in the exact same place two weeks later, days after all of the biomass
827 was consumed by the fire. Panel C is a closeup of the soil surface, showing in more detail
828 how the litter was also almost completely consumed by the fire. On the right, the photos in
829 panels d and e were on opposite sides of a fire line in an area that had minimal annual grass
830 invasion over a broad area, and thus lower fuel connectivity. Note the remaining plants and
831 stumps in panel e and the presence of only partially consumed litter in panel f.

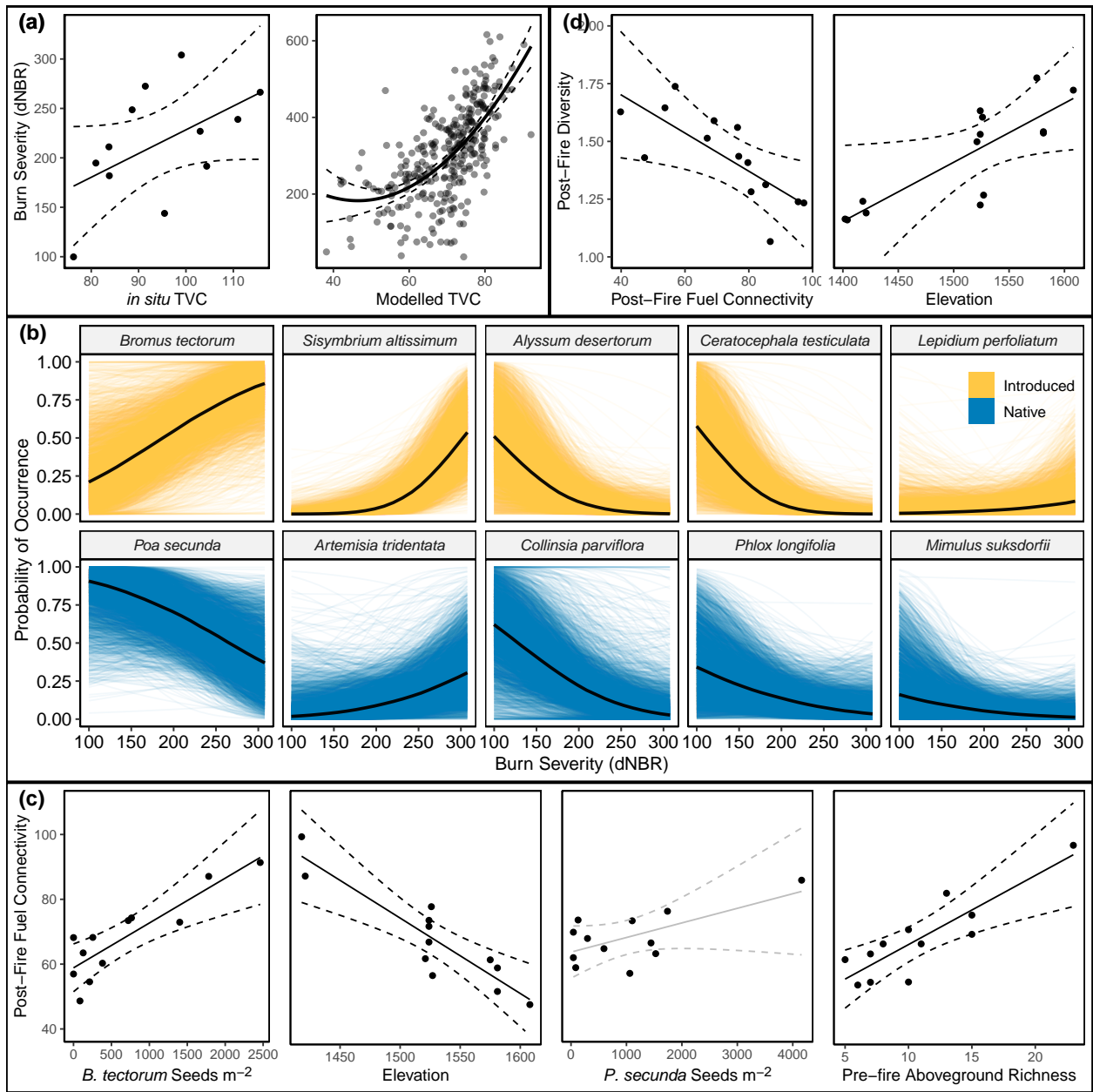


Figure 1: .

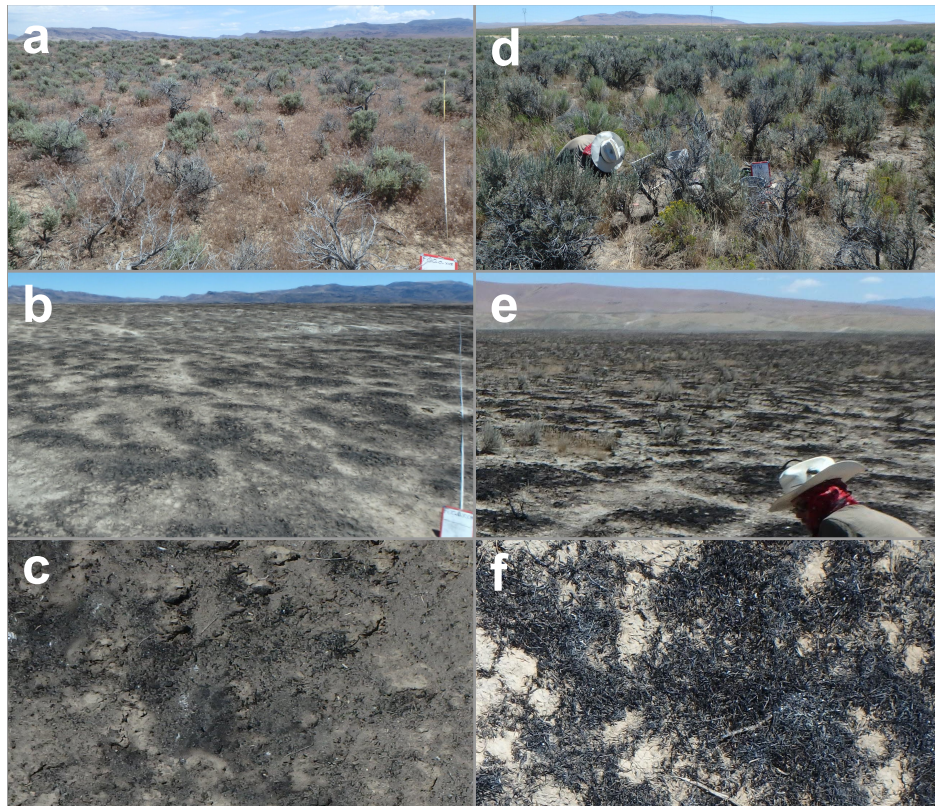


Figure 2: .