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1 **Evaluating pathways by which fuels and fires influence vegetation response**
2 **in a high-diversity plant community using structural equation models**

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17
18 **Running Headline: Above- and belowground fire effects on vegetation**

19 **Summary:**

20 **1.** Fire strongly influences plant populations and communities around the world, making it an
21 important agent of plant evolution. Fire influences vegetation through multiple pathways, both
22 above- and belowground. Few studies have yet attempted to tie these pathways together in a
23 mechanistic way through soil heating even though the importance of soil heating for plants in
24 fire-prone ecosystems is increasingly recognized.

25 **2.** Here we combine an experimental approach with structural equation modeling (SEM) to
26 simultaneously examine multiple pathways through which fire might influence herbaceous
27 vegetation. In a high-diversity longleaf pine groundcover community in Louisiana, USA, we
28 manipulated fine-fuel biomass and monitored the resulting fires with high-resolution
29 thermocouples placed in vertical profile above- and belowground.

30 **3.** We predicted that vegetation response to burning would be inversely related to fuel load
31 owing to relationships among fuels, fire temperature, duration, and soil heating.

32 **4.** We found that fuel manipulations altered fire properties and vegetation responses, of which
33 soil heating proved to be a highly accurate predictor. Fire duration acting through soil heating
34 was important for vegetation response in our SEMs, whereas fire temperature was not.

35 **5.** Our results indicate that in this herbaceous plant community, fire duration is a good predictor
36 of soil heating, and therefore, of vegetation response to fire. Soil heating may be the key
37 determinant of vegetation response to fire in ecosystems wherein plants persist by resprouting or
38 reseeded from soil-stored propagules.

39 **6. *Synthesis.*** Our SEMs demonstrate how the complex pathways through which fires influence
40 plant community structure and dynamics can be examined simultaneously. Comparative studies

41 of these pathways across different communities will provide important insights into the ecology,
42 evolution, and conservation of fire-prone ecosystems.

43

44 **Key-words:** disturbance, fire duration, fire temperature, first- and second-order fire effects,
45 longleaf pine savanna, residence time, resprouting, soil heating, structural equation modelling,
46 vegetation dynamics.

47

48 **Introduction**

49 Fire is an important evolutionary and ecological force that influences plant life in most
50 terrestrial ecosystems. As a potent agent of natural selection, fire shapes traits of plant species
51 and has likely done so since plants first colonized land (Bond & Keeley, 2005; Keeley & Rundel,
52 2005; Scott & Glasspool, 2006). As an environmental filter, fire often determines which plant
53 species occur within and dominate ecological communities (D'Antonio & Vitousek, 1992; Bond
54 & Keeley, 2005; Keeley & Rundel, 2005; Pausas & Verdú, 2008). Because fire is pervasive in
55 shaping vegetation structure and composition, and given the expectation that fire regimes will be
56 altered under global change (IPCC, 2007; Bowman *et al.*, 2009), we should strive to understand
57 the mechanisms by which fires influence plant populations and communities.

58 Fire influences vegetation through multiple, potentially interacting pathways that operate
59 both above- and belowground. Aboveground heat can kill plant tissue and sometimes individuals
60 outright. Although some trees can endure heat from fires, many fire-adapted plants persist by
61 resprouting from belowground organs or from seeds stored in the soil (e.g., Whelan, 1995;
62 Higgins, Bond & Trollope, 2000; Vesk & Westoby, 2004; Vesk, 2006). These organs and seeds
63 are susceptible to damage when fires on the surface heat the soil beyond some lethal time-

64 temperature threshold (e.g., temperatures above 60°C; e.g., Bradstock & Auld, 1995;
65 Choczynska & Johnson, 2009). Elevated soil temperatures are presumed to be a function of
66 aboveground fire temperature and duration (Steward, Peter & Richon, 1990; Bradstock & Auld,
67 1995). Because commonly used fire metrics are at best imperfect predictors of vegetation
68 responses (Keeley, 2009 and references therein), there is much we do not know about how fire
69 operates from a “plant’s eye view” (*sensu* Harper, 1977).

70 Despite widespread interest in the role of above- and belowground effects of fire on
71 plants (e.g., Keeley, 2009; Gagnon *et al.*, 2010 and references therein), empirical studies
72 commonly rely on snapshot-like aboveground fire metrics that can be poor predictors of
73 vegetation response. Such metrics include fire-line intensity, maximum fire temperature, and fire
74 severity (Johnson, 1992; Whelan, 1995; Bond & van Wilgen, 1996; Bond & Keeley, 2005). Fire
75 intensity refers to energy output during fire, whereas severity describes the amount of fuels
76 consumed (Keeley, 2009). These metrics are valuable for modeling fuels and behavior of fires,
77 but they can be poor indicators of damage to seed-banks and belowground plant organs, and
78 therefore, of longer-term population and community dynamics (Hodgkinson & Oxley, 1990;
79 Keeley, Brennan & Pfaff, 2008; Keeley, 2009). Such poor predictive power may be the result of
80 failure by these metrics to incorporate elements of soil heating and potential interactions of
81 above- and belowground processes on vegetation (Gagnon *et al.*, 2010). Given that many plant
82 species survive fires belowground (Vesk & Westoby, 2004; Vesk, 2006), fire metrics that
83 include some aspect of soil heating might better predict how fires affect plant populations and
84 communities.

85 Here we combine an experimental approach with structural equation modeling (SEM) to
86 examine above- and belowground pathways through which fires might influence vegetation. We

87 manipulated fine-fuel biomass to produce variation in fire properties, then measured fire duration
88 on the soil surface and temperatures in vertical profile. We developed hypotheses to explain how
89 above- and belowground fire properties might influence vegetation response, then used SEMs to
90 test the relative importance of multiple hypothesized pathways (Fig. 1) in a high-diversity
91 longleaf pine groundcover community in Louisiana, USA. Prior to prescribed fires, we
92 manipulated fuels and placed thermocouples at five different vertical positions. We predicted
93 that vegetation response would be inversely related to fuel load owing to complex relationships
94 among fuel load and fire properties above- and belowground. This prediction was validated, and
95 we found soil heating to be a highly accurate predictor of vegetation response. Our results
96 highlight the utility of SEMs for understanding complex, interrelated mechanisms through which
97 fires may influence the structure and dynamics of plant populations and communities.

98

99 **Methods**

100 *Study Site and Experiment*

101 We studied prescribed fires and their effects at Camp Whispering Pines (30° 41' N, 90°
102 29' W; 25-50 m.a.s.l.), a species-rich longleaf pine (*Pinus palustris* Mill.) savanna in
103 southeastern Louisiana, USA. Soils are Pleistocene-aged fine sands mixed with and capped by
104 loess, and are among the most fertile pine-savanna soils (McDaniel, 1990). When we began the
105 study, the site had been burned biennially during the early growing season (April-May) for the
106 previous 15 years (Noel, Platt & Moser, 1998). Additional site information is available in Platt *et*
107 *al.* (2006).

108 We manipulated fine fuels in our sample plots so that experimental fires would vary
109 substantially in temperature and duration. The first experimental treatment was increased-fuels,

110 in which we added 8 kg of dry, uncompacted longleaf pine needles, a highly flammable source of
111 fuel in this ecosystem (Fonda, 2001). All pine needles were dried and stored outdoors in plastic
112 bags under a rain shelter at the study site. We spread fuels evenly over the plots (each 2 x 2 m =
113 4-m²) on the same mornings as the two fires. This quantity of fine fuel (2 kg · m⁻²) mimicked the
114 upper range of observed fuel loads at this productive site (Thaxton & Platt, 2006). The second
115 treatment was reduced-fuels, in which we clipped and removed all biomass above 5 cm in height.
116 The third set of plots comprised unmanipulated control-fuels. We assigned these treatments
117 equally and randomly to 48 plots divided equally between two burn units (random blocks),
118 which we burned under prescription near mid-day on two different days. To reduce variability of
119 fuels among and within plots, we removed coarse woody fuels such as pinecones and downed
120 branches. We manipulated fuels immediately prior to lighting the fires. Following fuel
121 manipulations but before burning, we estimated total aboveground biomass by collecting all
122 biomass from a series of nearby plots to which the same three treatments were applied, and then
123 weighed the samples after drying for 48 hours at 100°C. Total aboveground biomass averaged
124 3076 g · m⁻² (± 57 g · m⁻² [1 SE]) in the increased-fuels treatment, 1076 g · m⁻² (± 57 g · m⁻²) in
125 the control, and 444 g · m⁻² (± 23 g · m⁻²) in the reduced-fuels treatment. These quantities
126 included natural herbaceous litter and any natural or added pine straw, plus naturally occurring
127 fine fuels like small pine twigs. Additional details of the experiment and a description of
128 bunchgrass responses to the fuel manipulations are in Gagnon *et al.* (2012).

129 To measure fire properties, we deployed high-resolution fire loggers at five positions in a
130 vertical profile (Grace, Owens & Allain, 2005; Ellair & Platt, 2013). We built the fire loggers
131 using HOBO® U12-014 J,K,S,T Thermocouple Data Loggers and Type K subminiature
132 connectors (Onset Computer Corporation USA), and Inconel 600-insulated (10') Type K

133 thermocouple wires (Omega Engineering, Inc. USA). We assembled the loggers and packaged
134 them in waterproof plastic containers, which we buried 10 cm below the soil surface outside the
135 sample plots on the morning of the fires (Grace, Owens & Allain, 2005). Although the data-
136 loggers were capable of recording temperatures from 0 to 1250°C with an accuracy of ± 4 °C
137 every second for 12 hours, the thermocouples to which they were attached were ultimately what
138 determined data-logger accuracy. Rather than measuring true flame temperatures, thermocouples
139 measure their own temperatures, which are subject to lags as a function of thermocouple
140 thickness (i.e., mass); accordingly, they systematically underrepresent true temperatures
141 (Kennard *et al.*, 2005, Wally, Menges & Weekley, 2006). Even so, their measurements are
142 comparatively accurate, albeit systematically biased, and are useful for regression analyses
143 (Kennard *et al.*, 2005) like those underpinning our SEMs. We located thermocouples at the soil
144 surface in all 48 plots, and in four other positions (1 cm above the soil surface and 1, 2, and 4-cm
145 below the soil surface) in 18 randomly selected plots (N = 6 plots/treatment; N = 3
146 plots/treatment/burn unit). We did so in a 1-m² sample quadrat in the center of each 4-m² plot on
147 the morning of the fires. For belowground measurements, we used a marked wooden dowel to
148 poke holes of appropriate diameter and depths, then inserted each thermocouple tip to the base of
149 the appropriate hole; we then sealed the soil around each protruding thermocouple cable by
150 lightly pressing the soil around it. In this way, we ensured that each thermocouple was buried to
151 appropriate depth with minimal soil disturbance. We secured thermocouples at the surface using
152 galvanized wire U-stakes ~3-cm from their tips. We additionally bent U-stakes into loops that
153 held thermocouple cables at 1-cm height.

154 We ignited prescribed fires during late morning on two dry days with light breezes in late
155 May 2007. We first set fires along the downwind perimeter of each of the two burn units; these

156 backing fires traveled into the wind. We then set head-fires along the upwind perimeter of each
157 burn unit; these burned through the plots in the direction of the wind. Reduced-fuel plots burned
158 with fine-scale patchiness, whereas control- and increased-fuel plots all burned thoroughly. Fuels
159 in all increased-fuel plots burned almost completely to ash. As fires in pine savannas burn
160 quickly (fires at the surface in our control plots averaged 10 sec. residence times), we were able
161 to remove even the belowground thermocouples from plots beginning 105 minutes after the fires.
162 Following the fires that afternoon, we used a leaf-blower and blew residual ash from all burned
163 plots. We collected and then replaced 0.5 kg of the ash on a random subset of plots; we found no
164 measurable effect of ash on vegetation response, so we do not consider ash further.

165

166 *Data Collection*

167 We calculated two fire metrics for the soil surface in each plot. Maximum temperature
168 increase was the difference between the hottest temperature during the fire and the ambient
169 temperature prior to the arrival of the flame front (Box 1). The second was fire duration, defined
170 as the time between when temperatures increased more than 0.3°C per second and the time they
171 fell below 50°C. In those few plots in which temperatures never exceeded 50°C, we instead used
172 the time following hottest temperature at which temperatures returned to within 5°C of pre-fire
173 ambient temperature. We calculated maximum temperature increase from every logger and fire
174 duration (i.e., residence time) from surface loggers only, using a custom R script.

175 We measured effects of fuel manipulations on vegetative cover in the 1-m² sample
176 quadrats within the center of the 4-m² fuel-treatment plots. We took photos 2 m above every plot
177 from a stepladder 3 weeks after the fires. By this time, *in situ* resprouting and some germination
178 was already occurring across the burned area, while post-fire germination of seeds arriving from

179 outside the plots was yet unlikely (Myers & Harms, 2011). Prior to fires, we inserted nails in
180 each 1-m² sample quadrat at 10-cm intervals, creating a grid of 100, 10x10-cm “cells” visible in
181 the photos. We counted the number of cells out of the 100 in each quadrat that contained any
182 green vegetation. This yielded a proportion of cells containing green vegetation as a measure of
183 short-term vegetation response. Prior to burning, this metric was 100% in all plots.

184 We examined effects of increased-fuels on post-fire germination from the soil seed-bank
185 in a concurrent experiment at the same study site (Table S1 in Supporting Information). We
186 applied two of the same fuel manipulations (control- and increased-fuels) to a separate set of
187 plots located in the same two burn units (see Myers & Harms, 2011 for details). In each of 60, 2
188 × 3 m plots (N = 30 increased-fuels, N = 30 controls), we collected a 20 × 20 × 1 cm (length ×
189 width × depth) soil sample (excluding litter) within one week after prescribed fires, which was
190 before most individuals began to germinate or resprout in the field. We sieved each soil sample
191 as described by Ter Heerd *et al.* (1996), spread each sieved sample thinly on top of sterilized
192 soil in individual trays, and monitored seedling emergence and species composition in a climate-
193 controlled growth chamber. We set light (16-h day length), temperature (32°C day, 22°C night),
194 and relative humidity (90% day, 50% night) to approximate growing-season conditions. We
195 watered and rotated trays regularly, recording abundance and species identity of germinating
196 plants for two months, by which time new seedling emergence had virtually ceased.

197 In both burn units we quantified effects of fuel manipulations on species presence in a
198 random subset of half the plots that contained surface fire-loggers. We identified all species with
199 aboveground living tissues (*e.g.*, stems, leaves) in the 24, 1-m² central quadrats during two pre-
200 fire censuses (conducted in July and October 2006) and two post-fire censuses (July and October
201 2007). We combined the 2006 censuses and combined the 2007 censuses because species were

202 often more readily identified during either summer or autumn. To compare and contrast both
203 species presence before and after the fires and relative patterns among functional groups, we
204 examined their frequencies of occurrence in quadrats among fuel treatments pre- and post-fire.

205

206 *Statistical models and analyses*

207 We used linear mixed-effect models to analyze fire temperatures, densities of plants
208 germinating from seed-bank samples, and species richness of seed-bank species. First we tested
209 for differences in hottest temperatures (b in Box 1) among the three fuel manipulations and five
210 vertical positions (Fig. 2, Table S2). For this analysis we used all 48 plots and fire loggers in all
211 five vertical positions in an unbalanced design. Based on quantile-quantile plots, box-plots and a
212 Shapiro-Wilk test, we log-transformed the response variable (hottest temperatures) to improve
213 normality and homoscedasticity and to eliminate overdispersion. A box-plot of the transformed
214 data and a Breusch-Pagan test both indicated heterogeneous variances, so we explored several
215 variance structures before grouping by fire logger position (Zuur *et al.*, 2009). After determining
216 the best-fit model using Akaike Information Criterion (AIC), we used post-hoc Tukey tests to
217 determine significance among treatment groups and their interactions. We tested for differences
218 in total species richness and in mean density of forbs and graminoids germinating from the seed-
219 bank using fuel treatments (control- and increased-fuels; N = 30 per treatment) as fixed effects
220 and blocks (burn units) as random effects (Myers and Harms, 2011). We performed all mixed
221 modeling in R (v.3.0.2) using the nlme package and the Tukey post-hoc comparisons using
222 lsmeans package (R Core Team, 2014).

223 We used linear regressions to explore relationships among fire temperatures, durations,
224 soil temperatures and vegetation response. We first examined proportion of cells containing

225 green vegetation as the response variable, which we logit-transformed using a 0.025 adjustment
226 factor to avoid 0 or 1 responses. Fire temperature, fire duration and soil temperatures (all log-
227 transformed) served as predictor variables (Fig. 3). We examined soil heating as the response to
228 fire temperature and duration at the soil surface, all log-transformed (Fig. S1). We performed
229 these regression analyses using the `lm` function in R (v.3.0.2) base package.

230 We built structural equation models (SEMs) to examine hypothesized pathways and
231 interactions through which fires on the surface might influence soil heating and vegetation
232 response. Construction of SEMs is guided by theory and *a priori* knowledge of the relevant
233 multivariate processes (including cause and effect) and is based on a series of bivariate
234 relationships among the various factors (Figs. 3 and S1). By evaluating such hypotheses using
235 SEMs, one can determine whether they are consistent with underlying patterns in the data. As
236 with any regression-based analysis, a concern with SEMs is an unfounded assumption of
237 causality among the proposed relationships, particularly when the data are observational. In this
238 study, relationships between fuel manipulations (our treatment) and temperature, duration, and
239 vegetation responses are all part of a controlled experiment. On the other hand, relationships
240 among surface and belowground fire properties and vegetation response are observational; these
241 we necessarily inferred from theory. We hypothesized that higher measured fire temperatures
242 and longer durations on the surface should increase belowground temperatures and reduce post-
243 fire resprouting and germination. Additionally, we hypothesized that increased fuels should
244 increase fire temperatures and durations.

245 Ideally we would have explored these hypotheses using a single SEM, but we were
246 constrained to building two separate models because of the limited size of our dataset of
247 belowground conditions. Our first model examined these relationships using our dataset of

248 surface conditions in all 48 plots (Fig. 1A). The diagram outlines our multivariate hypothesis
249 describing the effects of fuel manipulations on temperature and duration at the soil surface
250 during fire, and the combined effects of fuels, temperature and duration on vegetation response.
251 In the second model, we examined the role of belowground soil temperatures from the 18 plots
252 with fire loggers in vertical profile (Fig. 1B). We were unable to include fuel treatment in this
253 model because of our small sample size. Instead, we infer the effects of fuel treatment on
254 belowground temperatures from our mixed-model analysis (Fig. 2) and the results of the
255 aboveground SEM (Fig. 4A and B).

256 All data were not normal, so we applied transformations before conducting SEMs. To
257 correct for positive skew, we applied a natural log +1 transformation to above- and belowground
258 temperature-increase and fire duration. We applied a logit transformation to correct for strong
259 negative skew in vegetation response. All proposed relationships were linear following
260 transformations based on box-whisker plots and Shapiro-Wilk tests (from the UNIVARIATE
261 procedure in SAS release 9.3, SAS Institute Inc., Cary, North Carolina).

262 We included fuel treatments in the surface SEM as dummy-coded exogenous variables
263 (Fig. 1A). Control-fuel treatment does not appear in the diagrams because it serves as baseline.
264 The effects of increased- and reduced-fuel manipulations shown are in reference to this baseline.

265 To simplify the belowground model, we condensed the three measures of belowground
266 temperature-increase (i.e., at -1, -2 and -4 cm depths) into one composite variable. For this, we
267 used the first factor of a principal components analysis. This factor explained 95% of the
268 variation among the three variables; all three had a factor score > 0.97 .

269 We performed model estimation using maximum likelihood. We based model fit on chi-
270 square values and their associated *P*-values and judged a model as not fitting the underlying

271 structure in the data when it had a P -value < 0.05 based on a chi-square test. In the case of poor
272 model fit, we examined residual covariances, located the largest residuals, and added a model
273 pathway indicated by that residual. We did this only if the suggested pathway agreed with theory
274 and our understanding of the system. We deemed a model with a new pathway to be of value if it
275 satisfied a single degree of freedom chi-square test.

276 Path coefficients in our SEM figures indicate the strength of the various proposed effects
277 (arrows). These partial regression coefficients represent the change expected in an endogenous
278 variable if an exogenous variable is varied while the remaining exogenous variables remain
279 constant. We report both standardized coefficients (in standard deviation units) and
280 unstandardized coefficients. R^2 scores indicate the collective ability of the coefficients to explain
281 variation in the endogenous variables. Multiplying the relevant standardized path coefficients
282 indicates the strength of indirect effects.

283 To increase our confidence in the maximum-likelihood path coefficients, we conducted
284 two additional analyses. First, we addressed a concern that our dataset had low sample sizes
285 relative to the complexity of the models tested: for each model, the ratio (d) of sample size (n) to
286 the number of unknown parameters being tested (a) was < 7 . We therefore followed the
287 recommendation of Lee & Song (2004) for Bayesian estimation. This produced results virtually
288 identical to those of maximum likelihood estimation (for both models, path coefficients from
289 Bayesian estimation differed with those from maximum likelihood estimation by $< 1\%$). In our
290 second analysis, we accounted for a potential block (burn-unit) effect by including block in the
291 model as a dummy variable. We compared this model to one not including blocks and found no
292 significant effect of block (e.g., block added just 0.01 to the R^2 score of vegetation response).
293 Based on this result, we do not report results of models that included block. For all SEM

294 analyses based on maximum likelihood estimation, we used the lavaan package in R v.3.0.2
295 (Rosseel, 2012; Beaujean, 2014; R Core Team, 2014); for the Bayesian estimation, we used IBM
296 SPSS Amos version 20 as lavaan in R currently lacks this capacity (Arbuckle, 2011).

297

298 **Results**

299 *Mixed modeling of fuel treatment effects on above- and belowground temperatures*

300 Both fuel treatment and the position in vertical profile of thermocouples significantly
301 affected the hottest temperatures loggers recorded during fires. Of the five vertical positions we
302 examined, temperatures during fires were hotter by far at 1 cm aboveground and on the surface
303 than belowground (Fig. 2). On the soil surface, reduced-fuels produced the lowest measured
304 temperatures ($P < 0.001$ for reduced- vs. control-fuels at 0 cm; Tukey post-hoc tests), whereas
305 temperatures from control- and increased-fuels did not differ ($P = 0.141$). At 1 cm belowground,
306 mean hottest temperatures were only marginally hotter in increased-fuels relative to reduced-
307 fuels ($P = 0.059$). At both 2 and 4 cm belowground, the hottest temperatures were under
308 increased-fuels, whereas temperatures in control- and reduced-fuels were similar ($P < 0.001$
309 comparing increased- vs. control-fuels at both -2 and -4 cm; $P = 0.823$ and 0.801 comparing
310 control- vs. reduced-fuels, respectively). Only the increased-fuels treatment raised belowground
311 temperatures above 60°C – sometimes considered a lethal threshold – and not deeper than -2 cm.

312

313 *Effects of fuel treatments on species composition*

314 Increasing fuels reduced densities and species richness of seeds germinating from the soil
315 seed-bank after fires (Fig. 5). We identified 11 species in seed-bank samples, including 5 of forbs
316 (3 in the genus *Eupatorium*), 4 of C_3 grasses (all in the genus *Dichanthelium*), and 1 legume

317 (Table S1). Mean total densities of both forbs and graminoids were lower in increased-fuels plots
318 relative to control-fuels, and species richness was significantly reduced (Fig. 5).

319 Fires in increased-fuels also reduced occurrence of most species compared to control
320 plots based on plant censuses during the years before and after the fires (Table S3). With the
321 exception of some C₃ grasses, during the year after the fires, most species occurred less
322 frequently in increased-fuels plots than in control plots (Fig. S2). Several of the C₄ grasses
323 occurred less frequently in the increased-fuels plots. Strikingly, increasing fuel loads eliminated
324 over half of the forb species in the seed bank.

325

326 *SEM of aboveground influences of fire on vegetation response*

327 Our first SEM examined hypothesized relationships among fuel manipulations, fire
328 temperatures (i.e., maximum temperature increase at the surface), duration, and vegetation
329 response (see Figs. 3 and S1 for the bivariate relationships underlying this SEM and the next).
330 Maximum likelihood estimation of this model produced a chi-square of 47.56 with 2 df ($P <$
331 0.001), indicating that one or more important relationships in the data remained poorly described
332 (Fig. 4A). An examination of residual covariances revealed a strong unspecified relationship
333 between the increased-fuel treatment and vegetation response. A SEM that included this
334 relationship (Fig. 4B) had a chi-square value of 1.10, which was substantially lower than the
335 previous model, and easily passed the single degree of freedom chi-square test ($\Delta\chi^2 = 46.46 \gg$
336 3.841). Also, this model had a P -value of 0.295 (df = 1), indicating that it described the data
337 adequately to merit interpretation here.

338 Fuel manipulations had clear and strong relationships with both maximum temperature
339 increase and fire duration at the soil surface ($R^2 = 0.51$ and 0.62 respectively; Fig. 4B). Plots with

340 increased-fuels had hotter fires of longer duration than controls, whereas plots with reduced-fuels
341 had cooler fires with similar durations compared to controls. According to our thermocouples,
342 fire raised temperatures on the soil surface by an average of 361°, 216°, and 58°C and lasted an
343 average of 35, 10 and 8 seconds respectively in increased-, control- and reduced-fuels.

344 The proposed model indicated that vegetation cover was strongly reduced following fires
345 where we increased fuels and when fires at any given point lasted longer than 35 seconds (Fig. 3,
346 Fig. 4B). Our increased-fuel treatment had a large direct effect on vegetation response, reducing
347 it substantially (standardized path coefficient = -0.71). The second most important pathway was
348 that of increased fire duration (-0.25), which also suppressed vegetation response. The pathway
349 from temperature increase to vegetation response (-0.04) was not significant. For plots with
350 increased-fuels but low fire durations, some contained new green vegetation in fewer than half of
351 sampling cells, whereas others were revegetating more completely (Fig. 3). All control- and
352 reduced-fuels plots contained green vegetation in more than 90% of sampling cells, but increased
353 fire duration still caused a slight negative effect (Fig. 3 and Fig. 4B). In contrast to some direct
354 pathways, indirect pathways from fuel manipulations to vegetation response were all relatively
355 weak (e.g., the strongest was from increased-fuel treatment via duration at $0.71 \times -0.25 = -0.18$).
356

357 *SEM connecting fire aboveground to soil heating and vegetation response*

358 Our second SEM examined hypothesized relationships among fire temperature, duration,
359 belowground soil temperature, and vegetation response. Maximum likelihood estimation of this
360 model produced a chi-square of 2.75 with 2 df ($P = 0.25$), indicating that it described the data
361 adequately. The proposed model indicated that fire duration was strongly associated with soil
362 heating, whereas fire temperature at the surface was not (Fig. S1 and Fig. 4C). In turn, the model

363 indicated that soil heating was strongly and negatively associated with vegetation response. Fire
364 duration on the surface had a substantial indirect, negative association with vegetation response
365 ($0.63 \times -0.92 = -0.58$), whereas the indirect association between temperature increase at the
366 surface and vegetation response was weak ($0.14 \times -0.92 = -0.13$).

367

368 **Discussion**

369 Our structural equation models underscore the importance of fire duration operating
370 through soil heating as a determinant of herbaceous vegetation response to burning. Post-fire
371 resprouting and reseeded of herbs was strongly and negatively associated with shallow soil
372 heating, which was in turn strongly associated with fire duration (Fig 4C). By contrast,
373 aboveground maximum temperatures measured by thermocouples during fires were unimportant.
374 These observations are consistent with the hypothesis by Gagnon *et al.* (2010) that fires with
375 long residence times should send more heat into the ground and less upward into the air
376 compared with intense, fast-burning fires. But we caution that this study was not designed as a
377 test of that prediction and should not be interpreted as one; components of the study were
378 necessarily correlative, including relationships among fire properties and vegetation response.
379 Even so, the controlled experiment at the core of our study permits causal inferences about how
380 fuels influence both fire properties and vegetation response.

381 This study addresses the paucity of research linking herbaceous vegetation response to
382 fire, fuels and soil heating (as noted by Dickinson & Ryan, 2010; Stephan, Millar & Dickinson,
383 2010). In predicting vegetation response to fire, most previous studies have relied exclusively on
384 aboveground metrics (e.g., Johnson, 1992; Whelan, 1995; Bond & van Wilgen, 1996; Odion &
385 Davis, 2000; Bond & Keeley, 2005). Only a few studies have systematically examined the

386 effects of soil heating on herbaceous vegetation, and fewer still have attempted to
387 mechanistically link the effects of fire to response of herbaceous vegetation through soil heating
388 (e.g., Bradstock & Auld, 1995; Santana, Baeza & Blanes, 2013). Our findings are consistent with
389 the few other studies to have examined related questions. For example, Bova & Dickinson
390 (2005) found that fire residence time was a much better predictor than fire intensity of both heat
391 flux and depth of heating in tree trunks. Others have similarly concluded that fire temperatures
392 are not particularly useful for predicting effects of surface fires on soils (e.g., Van Wagner &
393 Methven, 1978, Bova & Dickinson, 2008).

394 Our short-term metric of herbaceous vegetation response is an accurate proxy for longer-
395 term effects on vegetation. In a related study from the same plots and fires, Gagnon *et al.*, (2012)
396 concluded that the increased-fuel treatment altered and suppressed the resprouting of individual
397 bunchgrass tussocks for the duration of the growing season. Similarly, Myers & Harms (2011)
398 monitored living, rooted plants in nearby plots at this same site after similar fuel manipulations
399 and found community-wide effects that persisted for at least two growing season. Given the
400 persistent effects we have documented elsewhere, it is likely that the reduced vegetative response
401 we detected three weeks after fires reflected substantial damage and mortality to plants in our
402 increased-fuel plots.

403 Surface fires typify our study ecosystem; fires that cause substantial soil heating reduce
404 the likelihood that individuals will survive to contribute to post-fire vegetation. This is generally
405 true regardless of a plants' species designation or functional group. Most plants we censused (>
406 90% of species) were herbaceous perennials that resprout to some degree; the large majority
407 persisted through surface fires in control plots. Since soil heating beneath increased-fuels
408 reduced overall vegetation cover, it is not surprising that frequency of occupancy generally

409 decreased as well. Although in some cases elevated soil temperatures can increase recruitment
410 from the soil seed-bank by triggering germination of fire-adapted seeds (e.g., Hodgkinson &
411 Oxley, 1990; Michaletz & Johnson, 2007), we found little evidence of that here. Instead, seed-
412 banking species produced a pattern similar to that of resprouting species, in that *per capita*
413 mortality increased under heavier fuel loads, with few obvious differences among species or
414 functional groups (Fig. 5, Fig S2). Given that many seedlings in our seed-bank study died before
415 growing large enough to identify, and because our seed-bank samples each came from a single
416 location in every sample plot, more extensive sampling of the seed-bank is needed to confirm
417 this result. A possible exception was a handful of C₃ grasses with higher frequencies after fires in
418 increased-fuels plots – one of these was *Panicum verrucosum*, a disturbance-tolerant annual.
419 Several C₄ grasses declined or were extirpated following fires in increased-fuels, a pattern
420 consistent with that reported by Gagnon *et al.* (2012) that bunchgrasses suffer under heavier fuel
421 loads. In this way, locally severe fires in heavy fuels may increase the availability of microsites
422 for colonization, a process that can influence spatial patterns of species diversity and community
423 composition in post-fire landscapes (Myers & Harms, 2011). These same conditions may also
424 increase abundances of disturbance-tolerant species (e.g., annual grasses), presumably owing to a
425 combination of heat-induced germination and higher plant performance in more open microsites.

426 Soil temperature of 60°C is sometimes considered the lethal threshold for plant tissues
427 (e.g., Bradstock & Auld, 1995; Choczynska & Johnson, 2009; but see Stephan, Miller &
428 Dickinson, 2010). In this study, only under increased-fuels did measured soil temperatures
429 exceed 60°C, and then not deeper than 2 cm belowground (Fig 2). Regardless, the reduced
430 resprouting and germination in these plots indicate that this admittedly simplistic threshold based
431 on thermocouple-measured temperature had merit for this system. Although various studies have

432 found dehydrated seeds surviving substantially hotter temperatures (e.g., Stephan, Miller &
433 Dickinson, 2010 and references therein), seeds in the soil of our study plots were killed by
434 temperatures measured around 60°C; that and the observation that our soils were moist suggest
435 that these seeds were hydrated and thus susceptible to the heat. Fire's influence on temperatures
436 declined quickly with soil depth, supporting the observation that soil is an excellent insulator
437 (Heyward, 1938; Beadle, 1940; Bradstock & Auld, 1995). That temperatures never approached
438 the lethal threshold in unmanipulated fuel-controls underscores the importance of heavy fuels
439 (e.g., downed branches, tree trunks and stumps) that burn for prolonged periods as gap-
440 producing hotspots in the groundcover that might serve as sites for post-fire colonization
441 (Thaxton & Platt, 2006; Myers & Harms, 2011; Wiggers *et al.*, 2013).

442 Our study suggests that fire duration and soil heating will be most useful for predicting
443 vegetation response in herbaceous, surface-fire systems like this longleaf pine savanna (Platt,
444 1999). Ecologists use many different measures of fire properties, and each is potentially useful
445 depending on the context (Keeley, 2009). In many ecosystems, maximum temperatures are
446 primarily a function of fine-fuel consumption (Beadle, 1940; Armour, Bunting &
447 Neuenschwander, 1984; Keeley, Brennan & Pfaff, 2008; Keeley, 2009), whereas fire duration
448 reflects the consumption of coarse or packed fuels (Hartford & Frandsen, 1992; Varner *et al.*,
449 2005; Michaletz & Johnson, 2007; Varner *et al.*, 2007, Varner *et al.*, 2009). The latter is more
450 likely to heat the soil (Gagnon *et al.*, 2010 and references therein; Massman, Frank & Mooney,
451 2010). We expect the relative importance of fire duration to increase additionally in ecosystems
452 where duff layers might alternately retain moisture and thus insulate the soil during relatively
453 brief fires, or dry out and heat the soil intensely when it combusts and smolders for prolonged

454 periods (Armour, Bunting & Neuenschwander, 1984; Hartford & Frandsen, 1992; Michaletz &
455 Johnson 2007; Varner *et al.*, 2007; Varner *et al.*, 2009; Butler & Dickinson, 2010).

456 Increasing fuel load had a substantial direct effect on vegetation response beyond
457 anything operating through fire temperature or duration (Fig. 4B). We view the most likely cause
458 as the difference in spatial scales between how we measured fire properties versus how we
459 measured vegetation response. Within our 1-m² sampling quadrats, we measured vegetation
460 response within one hundred small cells, whereas we measured fire metrics on the soil surface at
461 one single point per quadrat. Fire properties could vary greatly over very short distances because
462 fuels, and therefore combustion, were intrinsically spatially heterogeneous despite that we
463 specifically designed fuel manipulations to homogenize fire properties across the quadrat. Fuels
464 and fire temperatures, and thus thermocouple point measurements, are all inherently noisy at fine
465 spatial scales, and thermocouples are imperfect at best for measuring true flame temperatures
466 (Kennard *et al.*, 2005; Wally, Menges & Weekley, 2006). For similar studies in the future, we
467 recommend that researchers design tighter coupling of fire and vegetation metrics both in scale
468 and in space, for example by measuring fire metrics at multiple points within each sample plot
469 and then measuring vegetation response at those same points. We postulate that doing so here
470 would have produced a stronger effect of fuel treatment on vegetation via fire temperature and/or
471 duration, and a weaker direct effect of fuel manipulations (Fig. 4B). Alternatively, the direct
472 effect of increased-fuels on vegetation response may have been caused by an increase in ash and
473 accompanying soil nutrients on post-fire environmental conditions, residual aboveground
474 biomass, or biotic interactions (e.g., soil microbes, seed predators, plant competitors; Myers &
475 Harms, 2011; Gagnon *et al.*, 2012; Brown *et al.*, 2013). We view these as unlikely possibilities

476 because we applied fuel manipulations on the same days as burning and blew away ash
477 immediately afterwards, and subsequent ash manipulations had no effect on vegetation response.

478 An improvement to our method would be to use multiple, replicated plots across a broad
479 area, with each containing replicated thermocouple probes. Such a setup would enable data
480 capture during prescribed fires at broad scales but with high resolution both at the soil surface
481 and belowground; data could then be analyzed as we have done using SEM. By coupling this
482 setup with thermal imaging (e.g., Hiers *et al.*, 2009; Kremens, Dickinson & Bova, 2012),
483 scientists might partition fires into various constituent components (e.g., conductive, radiative,
484 and convective heat) to simultaneously compare the role of each on soil heating and subsequent
485 vegetation response. Data from such studies could inform predictive models of first- and second-
486 order fire effects (as per Dickinson & Ryan, 2010; Massman, Frank & Mooney, 2010; Stephan,
487 Millar & Dickinson, 2010) for the benefit of fire managers.

488 Our findings about the importance of fire duration relative to fire temperatures have
489 implications for conservation and management of both forests and herbaceous-dominated
490 systems. Soil heating is the key determinant of herbaceous vegetation response to fire in surface-
491 fire systems because those plants that persist through fires do so by resprouting from
492 belowground organs or by germinating from soil-stored seeds (Whelan, 1995; Higgins, Bond &
493 Trollope, 2000; Vesk & Westoby, 2004; Vesk, 2006). Thus, only by cooking their belowground
494 regenerative tissues are fires likely to kill plants outright (Flinn & Wein, 1977; Hodgkinson &
495 Oxley, 1990; Bradstock & Auld, 1995; Schimmel & Granstrom, 1996; Odion & Davis, 2000;
496 Brooks, 2002; Choczynska & Johnson, 2009; Gagnon *et al.*, 2010). Our results underscore the
497 need for extreme caution with dry, packed fuels that can smolder for prolonged periods at the
498 soil surface and thus heat the soil substantially (as per Varner *et al.*, 2009; Butler & Dickinson,

499 2010 and references therein). In addition, our results suggest that fire managers should consider
500 the advantages of fast-moving head fires that might cause less soil heating than creeping
501 backfires with longer residence times.

502

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509

510 **Data Accessibility**

511 - The complete dataset used for both ANOVA and SEM analyses: DRYAD entry doi:
512 xx.xxxx/dryad.xxxx.

513

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665

666 **SUPPORTING INFORMATION**

667 Additional supporting information may be found in the online version of this article:

668

669 **Table S1** Abundance and frequency of species emerging from soil seed-bank samples.

670 **Table S2** Results of mixed effects model of soil heating.

671 **Table S3** Species list, functional group classifications, and numbers of quadrats in which each
672 species was found, relative to prescribed fires and by fuel treatments.

673 **Figure S1** Bivariate regressions of the relationships between soil heating at 3 different depths
674 and fire temperature or duration at the soil surface on log-log scales.

675 **Figure S2** Comparison of frequency across quadrats of species found in increased-fuels
676 compared to control-fuels during the growing season after burning.

677

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680 typeset. Technical support issues arising from supporting information (other than missing files)
681 should be addressed to the authors.

682 **Figure Legends**

683 BOX 1. Diagram of a typical time-temperature series from a fire-logger located on the soil
684 surface including: (a) pre-fire ambient temperature, (b) hottest temperature, (c) maximum
685 temperature increase, (d) fire residence time, and total heat (shaded area).

686 FIG. 1. Hypothesized structural equation models of direct and indirect pathways: (A) from fuel
687 manipulations to surface fire temperature and duration to vegetation response, and (B)
688 from surface fire temperature and duration to soil heating to vegetation response. Circles
689 (e1-e3) signify error terms; double-headed arrows indicate significant correlations.

690 FIG. 2. Aboveground and belowground temperatures in the three fuel treatments. Boxes
691 represent the median and 25th/75th percentile. Whiskers extend to 1.5 times the
692 interquartile range. Letters above boxplots indicate statistical difference. Temperatures on
693 the Y-axis are log scale. Horizontal dotted line demarcates soil temperature of 60°C.
694 Vertical lines differentiate different depths; black line represents the soil surface.

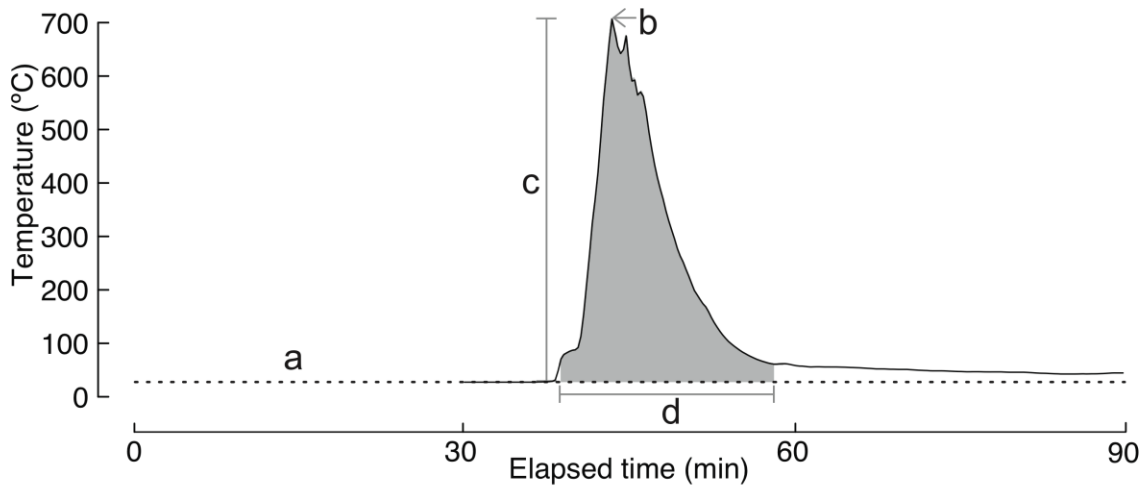
695 FIG. 3. Bivariate relationships between vegetation response and fire properties. The proportion
696 of cells containing green plants 3-weeks after burning represents vegetation response (Y
697 axes, on logit scale). Fire properties include temperature and duration on the surface and
698 temperature at three soil depths (X axes, on log scale) during experimental prescribed
699 fires. We incorporated these relationships into structural equation models. Black lines are
700 best-fit lines; gray areas encasing lines are 1SE envelopes.

701 FIG. 4. Structural equation models describing proposed relationships among fuels, fires and
702 vegetation. The models include: (A) our starting, theory-driven model describing
703 relationships aboveground ($\chi^2 = 47.56$, $df = 2$, $P < 0.001$); (B) the same model but with
704 an additional pathway from increased-fuels to vegetation response ($\chi^2 = 1.10$, $df = 1$, $P =$

705 0.295); and (C) our proposed model examining effects belowground ($\chi^2 = 2.75$, $df = 2$, P
706 $= 0.25$). Pathways are accompanied by standardized partial regression coefficients. The
707 significance of the coefficients is shown with differently weighted/colored lines (thin
708 gray = non-significant, medium black = $P \leq 0.01$, and thick black = $P \leq 0.001$). Models
709 in panels A and B have 48 samples, while the model in panel C has 18 samples. Circles
710 (e1-e3) signify error terms, double-headed arrows indicate significant correlations, and
711 R^2 values indicate the total variation explained by a model up to those points in the
712 diagram.

713 FIG. 5. Seed density and species richness from $20 \times 20 \times 1$ cm soil samples collected 1 week
714 after fires in control- and increased-fuels plots. Panels include: (A) density of forbs, (B)
715 density of graminoids, and (C) total species richness. Bars = back-transformed (density
716 only) least squares means ± 1 SE; $N = 30$. P -values from ANOVA are listed in panels.

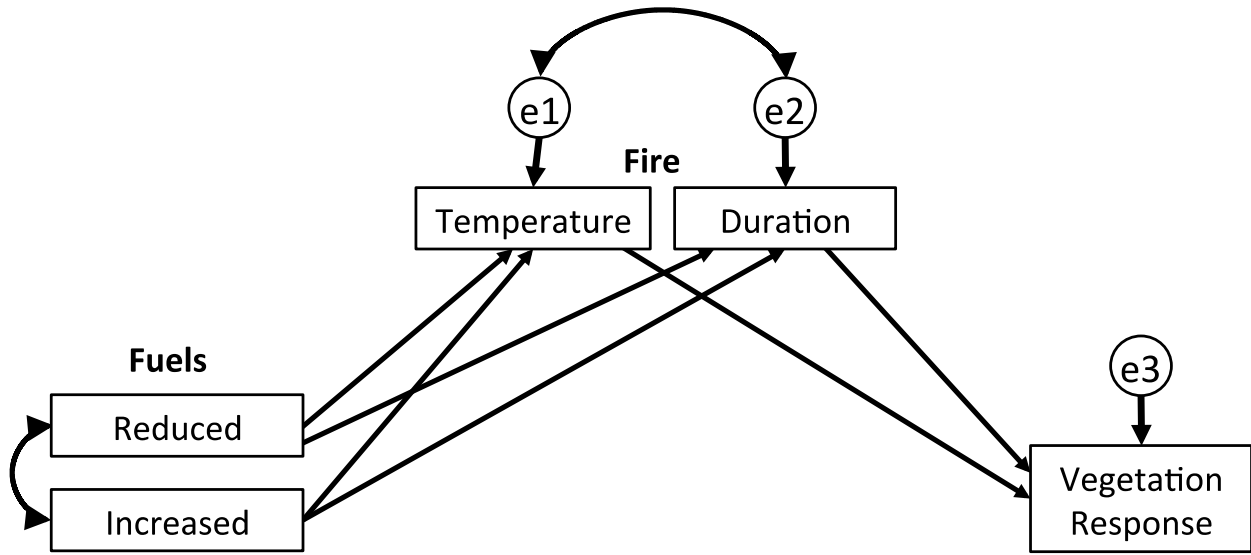
717



TERM	DEFINITION	UNITS
Ambient temperature	Average temperature prior to fire onset; (a) above.	°C
Hottest temperature	Highest temperature measured by a firelogger at a given location; (b) above. Locations were in vertical profile at soil surface and buried 1, 2, and 4 cm in the soil.	°C
Temperature increase	Difference between hottest fire temperature and pre-fire ambient temperature; (c) above.	°C
Fire duration	Time between fire onset and fire end as recorded by surface firelogger at given location; (d) above.	minutes:sec
Fire onset	First time temperature increased more than 0.3°C per second at given location	time of day
Fire end	Time at which temperature fell below 50°C at surface following fire at a given location	time of day
Fuel load	Quantity of combustible fuels at a given location comprising mostly grasses, herbs, pine needles and cones. Includes both fuels <i>in situ</i> and pine needles added experimentally in our fuel addition plots.	kg/m ²

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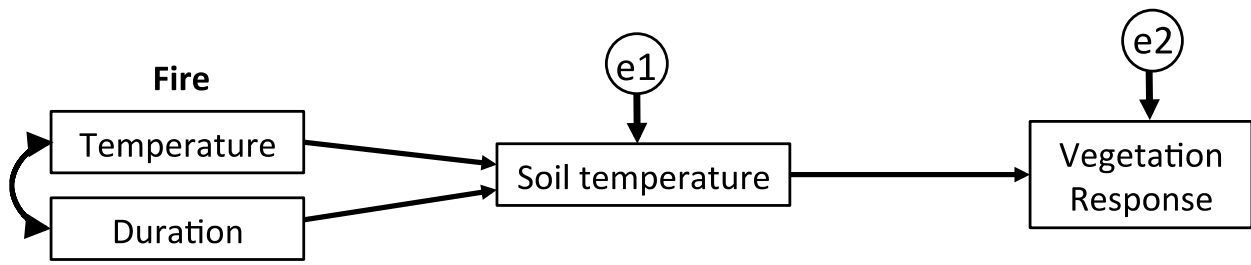
719 Box 1.



720

721 Figure 1A.

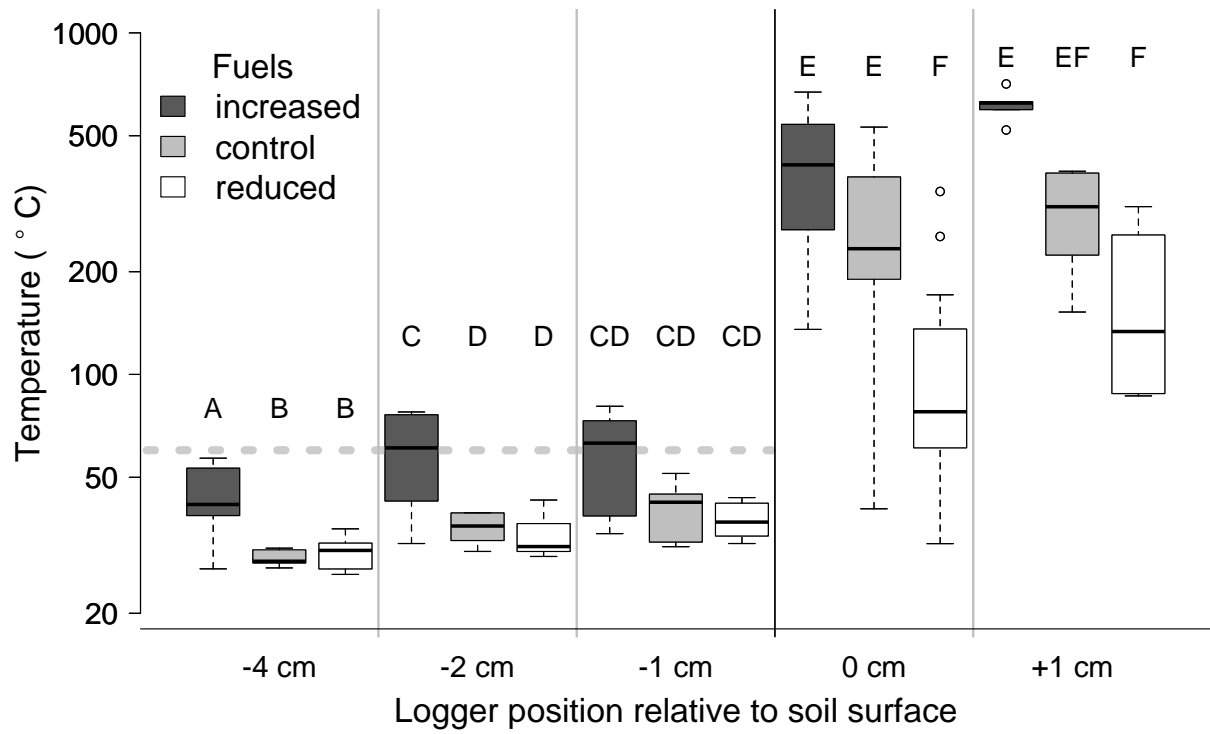
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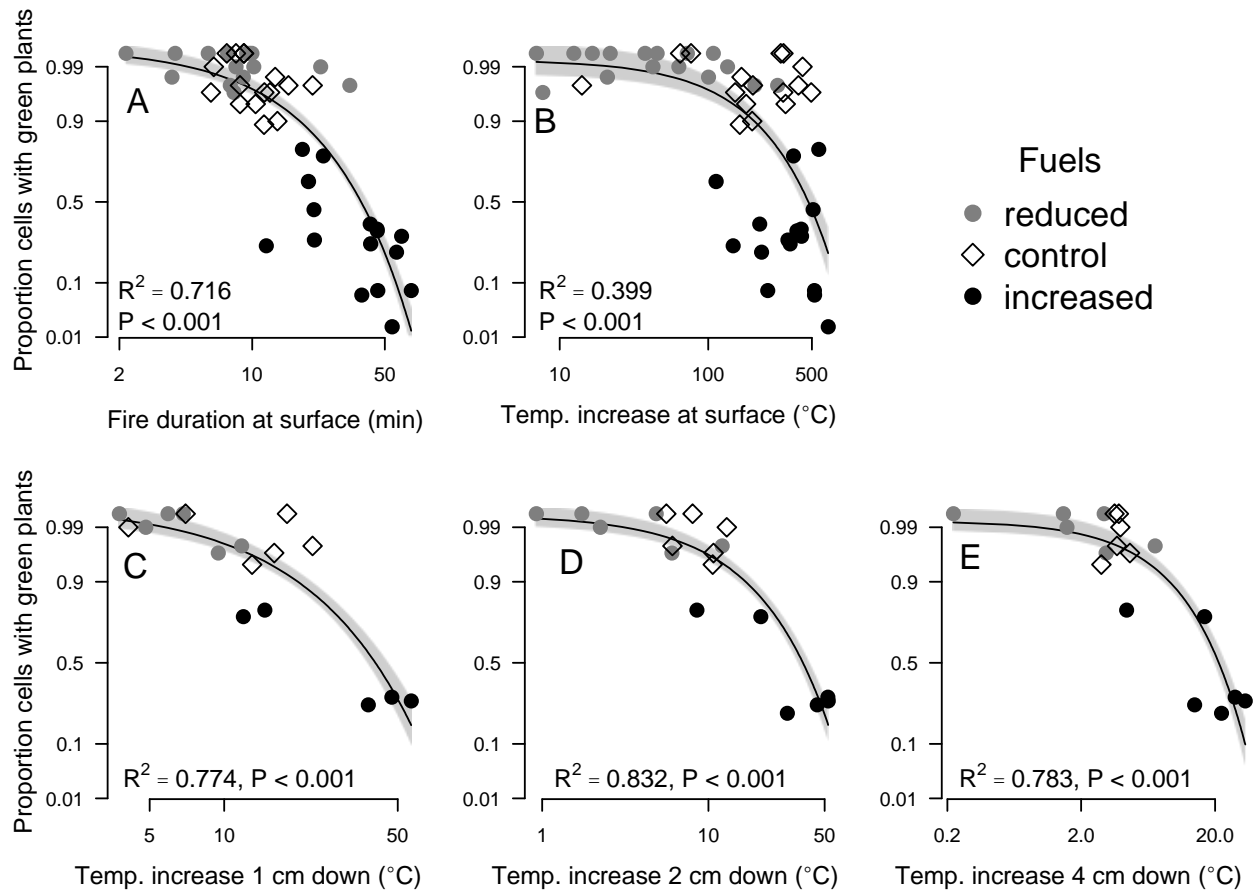
724 Figure 1B.

725



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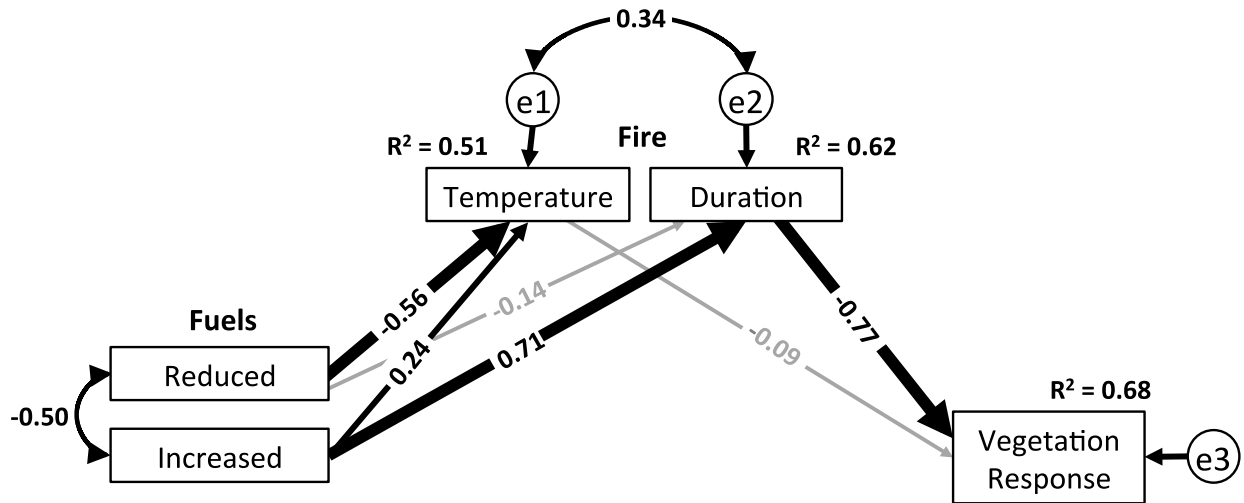
727 Figure 2.



728

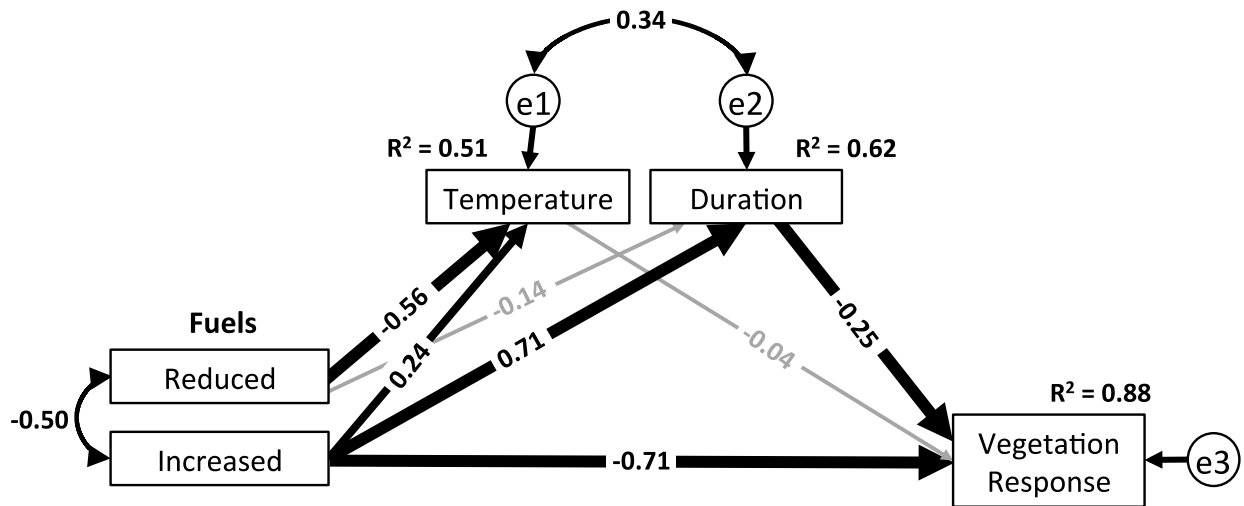
729 Figure 3.

730



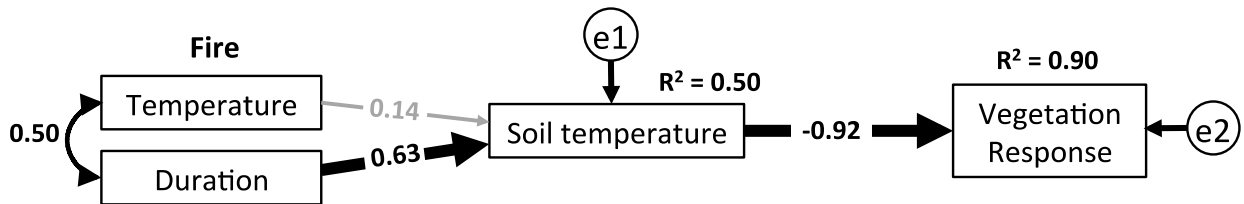
731

732 Figure 4A.



733

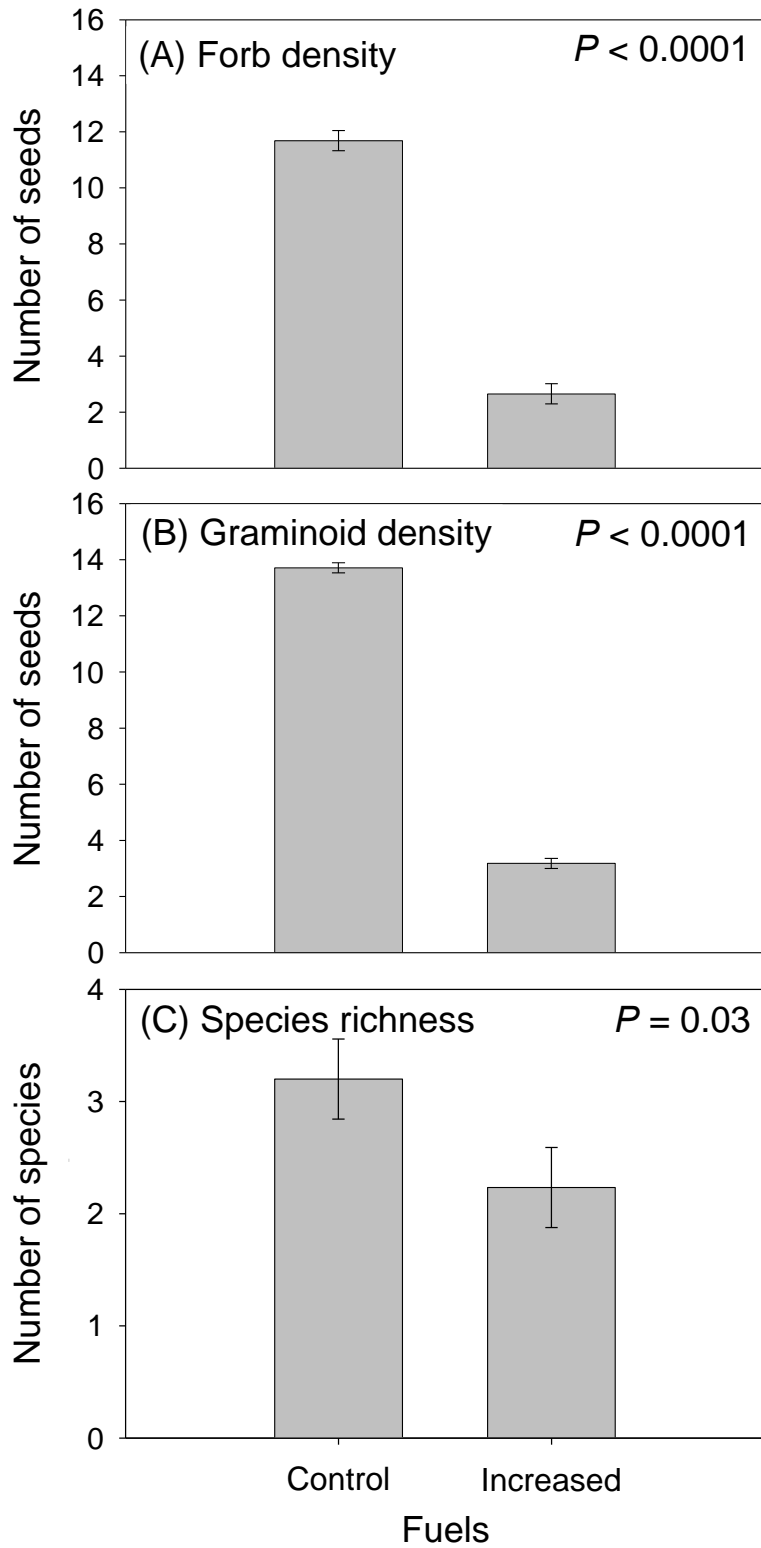
734 Figure 4B.



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736 Figure 4C.

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739 Figure 5.