Initial wildfire suppression efforts select for more extreme fuel and climate burning conditions in Sierra Nevada forests

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Abstract

Typical fire effects in Sierra yellow pine/mixed-conifer forests are dominated by large fires, and potential fire effects of rapidly suppressed fires are never realized. Milder fuel and weather conditions facilitate suppression, and thus typical fire effects materialize during more extreme conditions on average. This amounts to a selection pressure on burning conditions that may lead to a selection bias in average fire effects. We formalize this selection framework and measure its influence on correlated, multivariate burning conditions of fuel and climate using the evolutionary ecology concept of phenotypic selection. Using fire "survivorship" to initial attack suppression efforts as a fitness metric, we find that initial containment efforts select for fires burning in more homogeneous fuel conditions and during hotter/drier conditions. Fire effects arise from a complex social-ecological system, with management decision-making having a strong ability to influence outcomes. We show a strong selection pressure on burning conditions imposed by management, and encourage further dismantling of barriers to applying this selection for resource benefit, such as by expanding implementation of wildfire use fires where natural ignitions are allowed to burn under moderate conditions.

Introduction

A legacy of fire suppression is an oft-cited root cause underpinning the increasing size and severity of wildfires in the yellow pine/mixed-conifer forests of California's Sierra Nevada mountain range (Miller and Thode 2007; Calkin *et al.* 2015; Safford and Stevens 2017). While most of this ecosystem would experience frequent, low-to moderate-severity wildfire every 8 to 15 years in the several centuries prior to Euroamerican settlement, suppression management has largely eliminated fire effects from much of western dry forested land in the past

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100 years (Steel *et al.* 2015; Safford and Stevens 2017; Miller and Safford 2017). A lack of frequent fire has led to densification of Sierra yellow pine/mixed-conifer, which increases fuel loading and homogenizes forest structure (Collins *et al.* 2016; Stephens *et al.* 2018). Synergistic alignment of these extreme fuel conditions with earlier snow melt, longer fire seasons, and hotter droughts (aka "climate change droughts") (Westerling 2006, 2016; Abatzoglou and Kolden 2013; Abatzoglou and Williams 2016) increases the probability that fires will generate self-propagating behavior (Coen *et al.* 2018) and kill all (or nearly all) overstory vegetation (Koontz *et al.* 2019b) in large, contiguous patches of mortality (Stevens *et al.* 2017). The Sierra yellow pine/mixed-conifer community is ill-adapted to regenerate in the centers of these large patches, which are far from seed sources (Welch *et al.* 2016). Thus the modern trend of atypically large, contiguously stand-replacing fires in this system compromises forest resilience and increases the potential for long-term shifts in vegetation type to shrub- or grasslands (North *et al.* 2009; Millar and Stephenson 2015; Stevens *et al.* 2017).

Ongoing fire suppression also strongly influences fire effects. Fire suppression generally is very effective at its immediate goal of extinguishing fires; between 1970 and 2002, 97 to 99% of fires burning on U.S. Forest Service land were contained before they reached 120 hectares in size (Calkin *et al.* 2005, 2015). The "10 a.m. policy", which dictated that fires should be put out before 10 a.m. on the day following their discovery, has been modified since its establishment as firefighting policy in the 1930's but its aggressive spirit persists in today's modern firefighting apparatus (Dale 2006; Johnson *et al.* 2009). Mild weather and fuel conditions facilitate early suppression efforts (Calkin *et al.* 2015; Abatzoglou *et al.* 2018), and the fires that escape initial attack and grow large are often assumed to have grown to these sizes because they were burning under more extreme conditions (Calkin *et al.* 2014).

While the vast majority of fires are managed for suppression in the western U.S., including the Sierra yellow pine/mixed-conifer system (Calkin *et al.* 2015), a small number of natural ignitions are allowed to burn under moderate conditions as "wildland fire use" (WFU) fires in recognition of the benefit that fire has to forest resources (Davis 1979). WFU fire effects tend to fall within the natural range of variation for the Sierra Nevada yellow pine/mixed-conifer system (Meyer 2015). Though many studies recommend WFU fires as a means to restore forest resilience (Mallek *et al.* 2013; Meyer 2015; North *et al.* 2015; Collins *et al.* 2017; Stevens *et al.* 2017), barriers remain to their more widespread adoption (Doane *et al.* 2006).

The same mild fuel and weather conditions that contribute to beneficial outcomes of WFU fires contribute to early success of suppression efforts. Fire effects that may have arisen from fires that succumb to early suppression efforts never materialize. Instead, fire effects from suppression fires in this ecosystem are dominated by large fires, which account for approximately 97.5% of total burned area, exhibit fire effects that increase the likelihood of state change, and likely burn under the more extreme fuel and weather conditions

that hindered early suppression efforts (Calkin *et al.* 2005). Thus, the general short-term success of fire suppression policy paired with its long-term cumulative effect has led to a management paradox with respect to forest management aiming to restore resilience: *we shouldn't put out the fires that we can, and we can't put out the fires that we should.*

Wildfire suppression management has shifted the distribution of fire behavior, and therefore the distribution of fire effects, to be more extreme (Calkin *et al.* 2015). Here, we introduce the evolutionary ecology concept of "phenotypic selection" as a formal framework for measuring the magnitude of this distributional shift– the selection by suppression. We use a new dataset of fire severity (Koontz *et al.* 2019a) to quantify the strength of "selection" on wildfire burning conditions (regional climate, wind speed, vegetation density, vegetation continuity) imposed by initial attack suppression efforts, and discuss its implications for fire effects.

Methods

Study system

The Sierra Nevada yellow pine/mixed-conifer (hereafter Sierra YPMC) is a disturbance-prone forest system in the Sierra Nevada mountain range of California, U.S.A. It spans the full 628 kilometer latitudinal length of the Sierra Nevada, and 2,500 meters of elevation (300 meters to 2800 meters), primarily on the western slope of the mountain range (Safford and Stevens 2017). The forest is dominated by ponderosa pine (*Pinus ponderosa*), sugar pine (*Pinus lambertiana*), white fir (*Abies concolor*), and incense cedar (*Calocedrus decurrens*) in varying mixes. Prior to Euroamerican settlement, the ecosystem experienced frequent, low- to moderate-severity wildfire every 8 to 15 years on average (Steel *et al.* 2015; Safford and Stevens 2017), which consumed surface fuels but generally had minimal effects on large, established trees. This fire regime generated heterogeneous horizontal forest structure, with groups of relatively even-aged trees having interlocking crowns, individual trees with distinct crowns, and variably-sized gaps between these tree clump and individual tree structural features (Lydersen *et al.* 2013). A century of fire suppression has led to infill of these gaps, homogenizing the horizontal forest structure and compromising forest resilience in an era of climate change-induced hotter droughts (North *et al.* 2009; Millar and Stephenson 2015; Collins *et al.* 2016).

For our study, we compiled the Sierra YPMC type using the U.S. Forest Service Fire Return Interval Departure (FRID) dataset and included "dry mixed-conifer", "moist mixed-conifer", and "yellow pine" vegetation types following Steel *et al.* (2018) and Koontz *et al.* (2019b). These classifications represent "potential vegetation" given the climate of the area, such that there is no influence of recent disturbance events (Harvey *et al.* 2016; Steel *et al.* 2018; Koontz *et al.* 2019b).

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Context of Sierra YPMC wildfire

To describe the modern context of wildfire activity in the Sierra Nevada yellow pine/mixed-conifer system, we used geospatial records contained in Short (2017) (U.S. Forest Service Fire Program Analysis Fire Occurrence Database; FPAFOD), the most comprehensive database of wildfire occurrence for the United States representing 1.88 million wildfire records from 1992 to 2015.

The FPAFOD contains point locations for the centroids of each fire's footprint, rather than the perimeter of each fire as in some other databases (Eidenshink *et al.* 2007). We spatially subsetted the FOD data to fire events whose centroids occurred within the Sierra Nevada mountain range, as defined by the Jepson geographic subdivisions (north, central, and south Sierra Nevada Foothills and High Sierra Nevada, as well as the Tehachapi Mountain Area) (JepsonFloraProject 2016). For each fire record, we approximated its footprint by creating a circular buffer around the centroid with an area equivalent to the reported area of the fire. Using this footprint approximation, we calculated the proportion of area that intersected with our compilation of Sierra YPMC from the FRID dataset. We retained all fires with greater than zero area of the approximate footprint covering the Sierra YPMC extent. We calculated burn duration as the number of days between the containment date and the alarm date, and retained all fires with a burn duration of greater than 0 and less than 364 days to eliminate likely errors in reporting of alarm and containment dates (e.g., switched alarm date and containment date creating negative burn duration, typo in containment year creates a 5+ year burn duration).

Measuring wildfire severity

The CalFire Fire Resource and Protection Program (FRAP; http://frap.fire.ca.gov/) maintains the most comprehensive Dataset of wildfire perimeters in the state of California, including attribute data for each fire such as its discovery date, its containment date, and the management objective. The management objective represents the approach taken by the management unit overseeing the wildfire– either "suppression", with a goal of extinguishing the fire, or "wildland fire use", with a goal of allowing the fire to burn to benefit forest resources so long as it didn't threaten lives or property (Meyer 2015). This dichotomy is somewhat simplistic, as each wildfire can be managed for multiple objectives, but it is a generally useful framework for understanding the primary management goal (Meyer 2015). This dataset contains all fires >4 hectares (and a non-comprehensive set of fires <4 hectares), and thus has greater representation of fire events compared to other wildfire events datasets, though it lacks severity information. For instance, the Monitoring Trends in Burn Severity (MTBS) database only contains wildfires in the western U.S. that are larger than 400 hectares (Eidenshink *et al.* 2007) and the U.S. Forest Service Region 5 geospatial database contains wildfires in the

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122 Sierra Nevada that are larger than 80 hectares (Steel *et al.* 2018), though both of these datasets contain
123 information on wildfire severity.

124 Koontz *et al.* (2019a) used the expanded FRAP dataset of over 1,000 wildfire perimeters to calculate wildfire
125 severity and calibrate satellite-derived measures using ground-based overstory composite burn index (CBI)
126 (Koontz *et al.* 2019b), which is an integrated measure of the effect of a wildfire on the forest overstory
127 one year after the burn (Key and Benson 2006). CBI correlates well with direct measures of fire impact to
128 vegetation in Sierra YPMC, such as percent of overstory mortality (Miller and Thode 2007). Thresholds of
129 wildfire severity (unchanged, low, moderate, and high) calibrated to the full dataset (Koontz *et al.* 2019b)
130 were imposed on each fire and then contiguous pixels of each category were vectorized into polygons to form
131 patches of each severity category. We subsetted this FRAP-derived dataset of fire severity to 622 that burned
132 in majority yellow pine/mixed-conifer forest (547 suppression fires; 75 WFU fires).

Burning conditions per fire event

134 In addition to mapping wildfire severity across each fire in the FRAP perimeter database, Koontz *et al.*
135 (2019b) also calculated fuel and regional climate variables within the burn perimeter. The prefire Normalized
136 Difference Vegetation Index (NDVI; Rouse *et al.* (1973)]) was found to correlate strongly with local wildfire
137 severity, as was the standard deviation of NDVI within the 90m x 90m window surrounding each pixel,
138 which represents a measure of horizontal forest structure and fuel continuity (Koontz *et al.* 2019b). The
139 gridMET product (Abatzoglou 2013) was used to calculate the energy release component, a modeled estimate
140 of expected fire behavior in conifer forest, for the 3 days prior to each fire's discovery date as well as the
141 wind speed for the first three days of the fire. The gridMET product has a daily temporal resolution and a 4
142 kilometer spatial resolution, so our climate variables capture regional conditions over time periods of several
143 days, but not very local weather events that might occur over the span of hours. Each of these variables has
144 a strong impact on wildfire behavior at macroscales (Abatzoglou and Kolden 2013).

145 For this study, we assigned prefire burning conditions for each fire as the mean fuel (prefire NDVI, prefire
146 standard deviation of NDVI within 90m x 90m moving windows) and regional climate (energy release
147 component for 3 days prior to the fire, wind speed for first 3 days of the fire) values within each fire perimeter.

148 For each fire, we calculated the total number of fires burning on that fire's alarm date, the proportion of area
149 represented by each severity category, as well as the maximum patch size of each severity category. Finally,
150 we calculated the stand replacing decay coefficient (SDC) (Collins *et al.* 2017; Stevens *et al.* 2017), a single
151 metric that integrates high severity patch size and shape such that a lower SDC corresponds to a larger, more
152 circular high severity patches with effectively more area within those patches far from the patch edges.

Designating "survivorship" of suppression fires

For each wildfire in the Koontz *et al.* (2019b) dataset with a suppression management objective, we determined whether it "survived" initial attack by whether its burn duration (discovery date subtracted from the containment date) was greater than 1 day. Following Abatzoglou *et al.* (2018), we assumed that fires under a suppression management objective that burned for more than one day would require different firefighting tactics than direct attack and thus represented a reconfiguration of firefighting personnel and resource allocation.

Quantifying the selection effect of fire suppression

We treated the prefire fuel (prefire NDVI, heterogeneity of NDVI) and climate conditions (prefire energy release component, early fire wind speed) as wildfire "phenotypes" having some distribution, and used logistic regression to measure the extent to which wildfire suppression "selected" for particular burning condition phenotypes using the survivorship of each fire to initial attack as a binary response fitness metric in an evolutionary ecology framework (Lande and Arnold 1983; Janzen and Stern 1998).

Implications of selection by suppression

We graphically compare fire effects (event size, stand replacing decay coefficient, proportion high severity) of suppression fires to those of wildfire use (WFU) fires to draw inferences about how the "lost contribution" of fires that succumbed to early containment may have influenced fire effects.

Software and data availability

Our selection model was fit using the `brms` (Bürkner 2017) package in `R` (R Core Team 2018). Our workflow greatly benefited from the `tidyverse` group of packages (Wickham 2017), and we manipulated spatial data using the following `R` packages: `sf` (Pebesma 2018), `raster` (Hijmans *et al.* 2019), `velox` (Hunziker 2017), `stars` (Pebesma 2019a), `lwgeom` (Pebesma 2019b), `fasterize` (Ross 2018), and `APfun` (Plowright 2018).

The original severity dataset created by (Koontz *et al.* 2019a) can be found on the Open Science Framework (https://osf.io/ke4qj/). The data and code for this study can also be found on the Open Science Framework.

Results

Fire event size and burn duration context of Sierra YPMC wildfire

We found that 16219 fire events burned at least partially in Sierra Nevada yellow pine/mixed-conifer between 1992 and 2015 covering 2.19 million total hectares and 1.18 million hectares of this forest type. A total of 14873 fires burned in greater than 50% Sierra yellow pine/mixed-conifer in the same period. The vast majority of these fire events were very small. Comparing the distribution of fire sizes in this system between 1992 and 2015 to relevant reference sizes:

- 61.46% of fires were smaller than 0.09 hectares– the size of a single pixel from Landsat which is a USGS satellite product often used to measure fire effects by comparing imagery just before the fire to imagery one year after the fire (Miller and Thode 2007).
- 98.87% of fires were smaller than 400 hectares– the minimum fire size for inclusion in the MTBS dataset for the western U.S. (Eidenshink *et al.* 2007), meaning that MTBS would include approximately 1.13% of fires in this system.
- 97.93% of fires were smaller than 80 hectares– the minimum fire size for inclusion in the USFS Region 5 geospatial dataset (Steel *et al.* 2018), meaning that the USFS Region 5 data accounts for approximately 2.07% of fires in this system.
- 94.88% of fires were smaller than 4 hectares– the minimum fire size for inclusion in the CalFire FRAP fire perimeter dataset which was used to derive the Koontz *et al.* (2019a) severity dataset. Thus, the Koontz *et al.* (2019a) dataset includes approximately 5.12% of fires in this system during its time span of 1984 to 2017.

Context of suppression fires compared to wildfire use (WFU) fires

There is a clear multimodality in the distribution of suppression fire event size while wildfire use fire event sizes are lognormally distributed (Figure 1). Under a suppression management objective, most fires are successfully contained at relatively small sizes while WFU fires are larger on average. This effect is obscured when comparing median fire sizes between management objectives using just larger fires (Table 1).
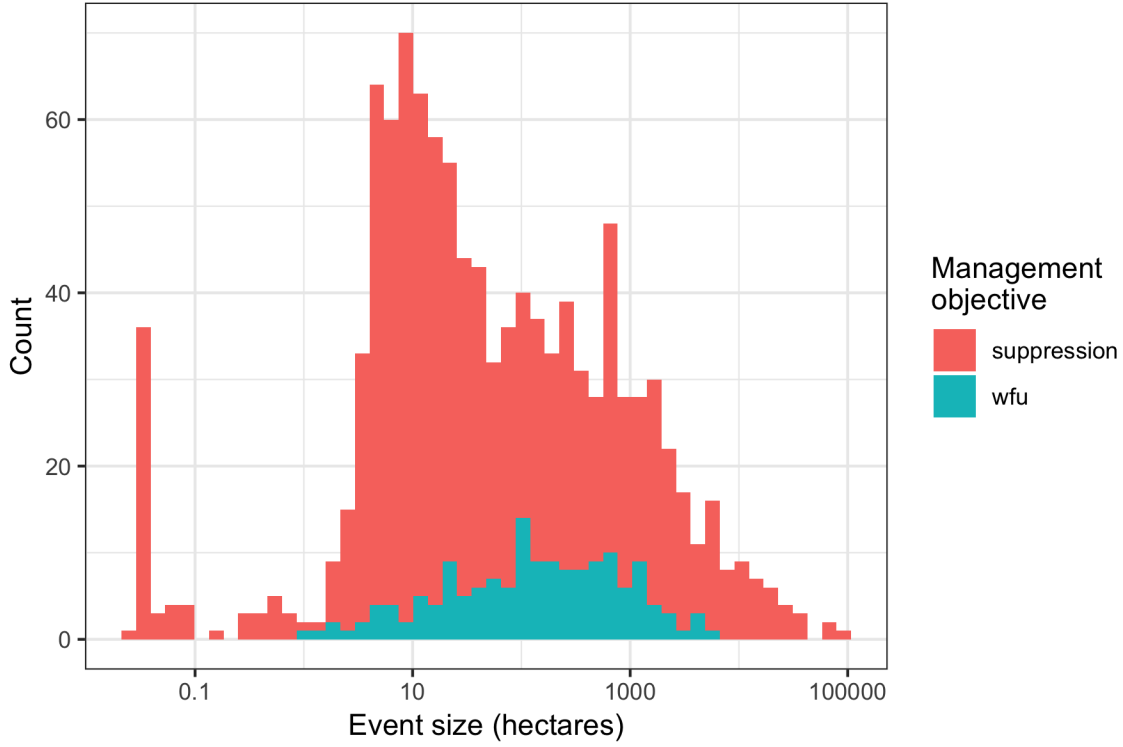
Figure 2: Distribution of log fire event size by management objective. While wildfire use fires exhibit a lognormal distribution in size, suppression fires exhibit clear multimodality with many fires extinguished when they are very small.

Table 1: Comparison of fire event sizes by management objective for fires burning in majority yellow pine/mixed-conifer in the Sierra Nevada between 1984 and 2017. Median fire size of all suppression fires is much lower than the median fire size of WFU fires, owing to the highly successful early suppression efforts. This size difference becomes less apparent as the minimum fire size in the dataset is increased, as is the case for fire effects datasets that focus on tracking larger fire events (e.g., U.S. Forest Service Region 5 GIS dataset with a minimum fire size of 80 ha or MTBS with a minimum fire size of 400 ha in the western U.S. The size difference is likely to be even more dramatic than what can be seen in the FRAP-derived Koontz *et al.* (2019b) dataset with its comprehansive coverage of fire greater than 4 ha (and some additional fires smaller than this threshold), as nearly 95% of wildfires in the Sierra yellow pine/mixed-conifer region remain even smaller than 4 ha.

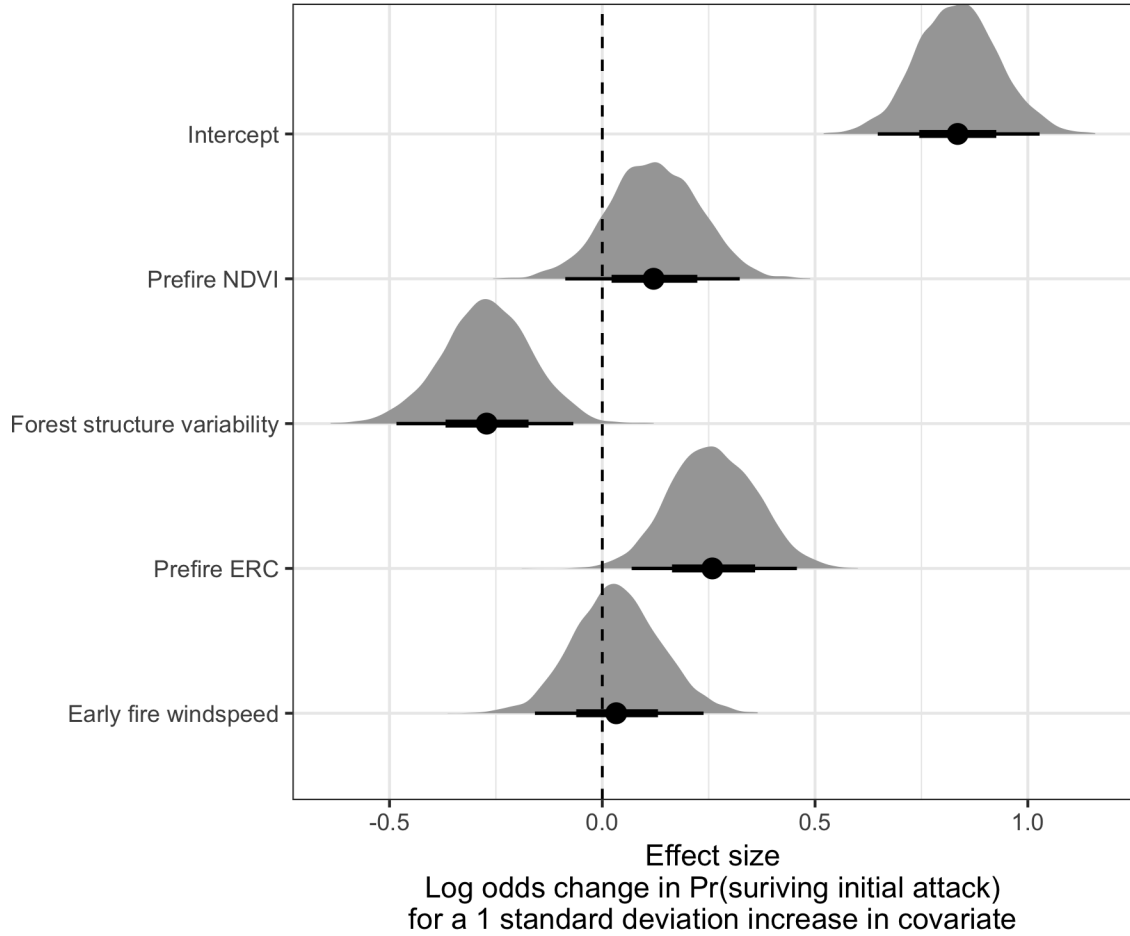| Management objective	Median fire size (ha) All FRAP fires	Median fire size (ha) Fires > 4ha	Median fire size (ha) Fires > 80ha	Median fire size (ha) Fires > 400ha
suppression	16.77	29.13	582	1459
wfu	193.1	205.4	506.7	866.5

Figure 3: Halfeye plot showing posterior distributions of coefficient estimates for model predicting the probability of wildfire survivorship in the first 48 hours of initial attack. The effect sizes are proportional to the 'strength of selection' of initial attack on the burning conditions of wildfire. Credible intervals are shown below each probability density function with the point representing the mean, the dark line representing the 66% credible interval, and the light line representing the 95% credible interval. The dotted line shows an effect size of zero.

Selection by suppression

We found a sizable effect of phenotypic selection of initial attack suppression on the heterogeneity of NDVI and prefire energy release component. Wildfires that survived initial attack were more likely to be burning in more extreme, homogeneous fuels (lower heterogeneity of NDVI) and in more extreme, hotter/drier climate conditions (higher energy release component) compared to wildfires that succumbed to the initial attack. We detected no selection pressure on prefire vegetation density as measured by NDVI or wind speed during the first three days of the fire (Figure 3; Figure 4).
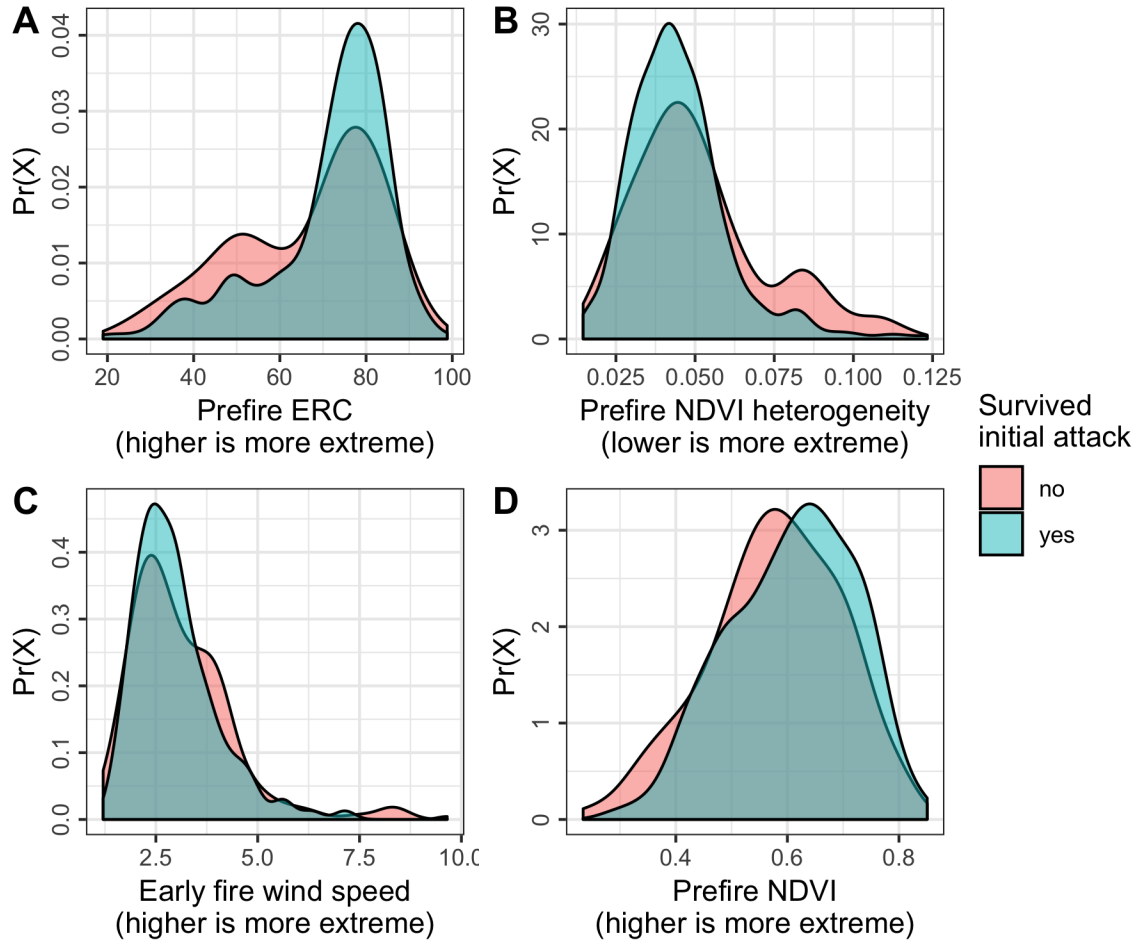
Figure 4: The selection effect on the vegetation and climate burning conditions of suppressed wildfires. A) Early suppression efforts selected for greater average energy release component (ERC), an estimate of fireline intensity correlated to hot, dry conditions. B) Fires surviving initial suppression efforts burn in more homogenous fuels, as measured by the standard deviation of NDVI in a 90m x 90m window. C) We detected no selection pressure on windspeed for the first 3 days of the fire. D) We detected no selection pressure on prefire NDVI, which is correlated with overstory canopy density and surface fuel loads.

Implications of selection by suppression for fire effects

We also found that fire effects of event size, proportion of high severity, and stand replacing decay coefficient showed a similar trend (i.e., slope) in suppression versus WFU fires as a function of burn duration, except for those suppression fires with short burn durations. However, the average fire effects (i.e., intercept) of suppression fires tended to depart from effects of WFU fires and resulted in greater proportion of high severity and a lower stand replacing decay coefficient. The notable exception to these trends are suppression fires with short burn durations, which experienced a strong selection pressure on their burning conditions and whose fire effects were dramatically different from fire effects of suppression fires that burned for longer. Suppression fires that succumbed early to containment were much smaller, had lower proportions of high severity, and had much higher stand replacing decay coefficients than suppression fires that escaped initial attack Figure 5). The fire effects of suppression fires that were successfully contained were within the range of variation of WFU fires for similar burn durations.

Discussion

Wildfire effects to forest vegetation are outcomes of a complex social-ecological system dynamic (Calkin *et al.* 2015). Direct causes of fire effects to vegetation arise from fire behavior and intensity (Keeley 2009), which are coupled to fuel, weather, and topography (McKenzie and Hessl 2008; Cansler and McKenzie 2014; Harvey *et al.* 2016). Wildfire effects are indirectly related to firefighting resource availability, legacies of management policy that change fuel distributions, and incentive structures that often prioritize mitigating short term loss of resources over long-term benefits of wildfire (Houtman *et al.* 2013; Calkin *et al.* 2014, 2015). Fire suppression allows some fires to "survive" initial containment efforts and contribute to fire effects on the landscape, while other fires are "killed" by initial containment and their potential fire effects are never realized. Because milder fuel and climate conditions facilitate fire containment efforts, initial attack suppression imposes selection for fires that burn under more extreme conditions and shifts the concomitant distribution of fire effects to also be more extreme. We measured this selection pressure and found a sizable influence of initial suppression efforts that decreased average fuel heterogeneity and increased average energy release component for fires that contribute to fire effects on the landscape (i.e., those that survive initial containment). Two primary implications follow from our findings:

1) There is a lost contribution to overall fire effects in Sierra Nevada yellow pine/mixed-conifer forests associated with fires that are extinguished before they have a chance to burn under the milder conditions that facilitated their early containment (Figure 5).
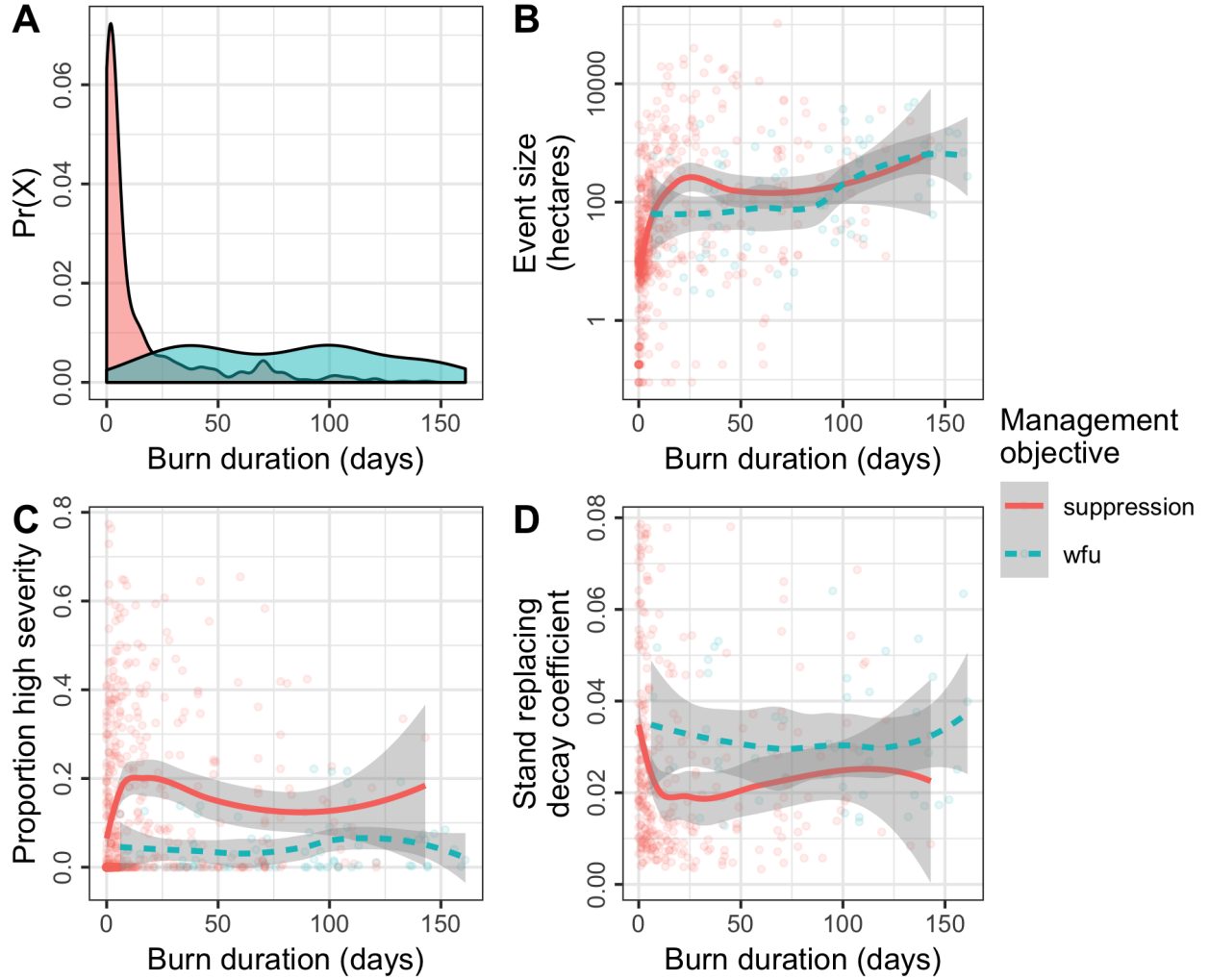
Figure 5: A) Distribution of burn duration by management objective. Most suppression fires are quickly extinguished. B) Effect of burn duration on fire event size shows that there's a similar trajectory between suppressed and wildfire use fires except early in the burning period when suppression fires remain small. C) The high severity portion of the fire tends to increase with shorter-duration suppression fires, but is relatively constant across burn durations for wildfire use fires. D) For fires with a high severity component, the stand replacing decay coefficient sharply declines for short-lived fires indicating that the high severity patches are larger and simpler which may reduce tree propagule pressure in the center of these patches and compromise dry forest regeneration. The SDC tends to increase with the burn duration for wildfire use fires.
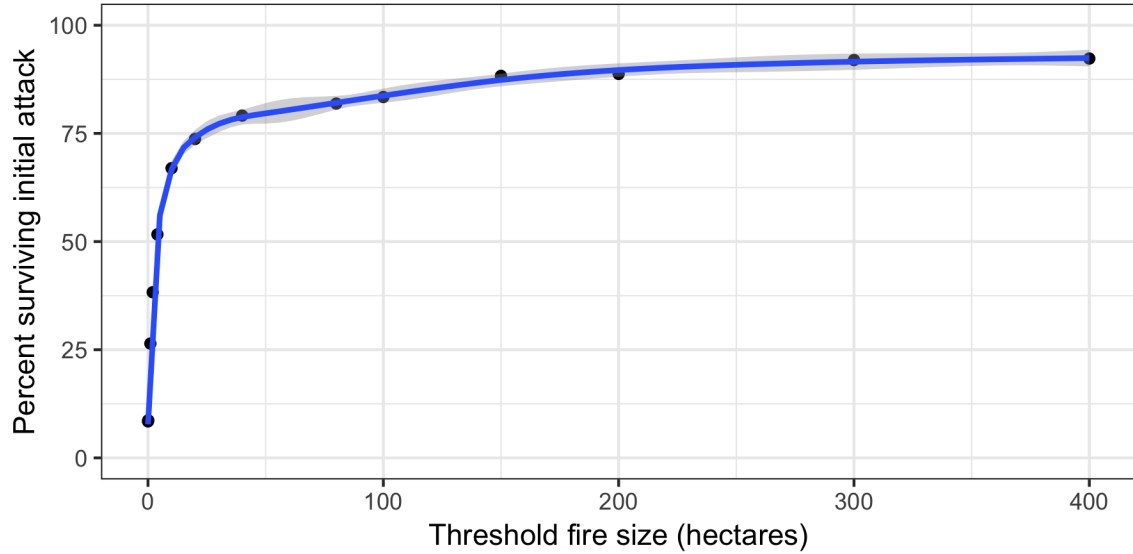
Figure 6: As the minimum fire event size of a dataset increases, a greater proportion of those fire events survived initial attack suppression efforts and burned on average under more extreme conditions. Thus databases with larger minimum fire sizes exhibit a stronger bias towards fires that burned in extreme conditions as a result of selection by suppression.

239   2) The fire effects that are measured only on large fires reflect a selection bias imposed by suppression

240   efforts (Figure 6).

241 **Context of suppression management on fire effects**

242 We compared fires managed for suppression to those managed for wildfire use/resource benefit to assess how

243 the selection pressure on burning conditions may ultimately influence fire effects. We demonstrated a clear

244 multi-modality in the fire event size distribution of wildfires managed with a suppression objective (Figure 1).

245 The median fire size of all fires in the dataset was much larger for WFU fires than for suppression fires, but

246 the median fire size of *large* fires (e.g., fires greater than 80 or 400 hectares) was much larger for suppression

247 fires (Table 1). Miller and Safford (2017) found a similar pattern in comparing the average size of modern

248 fires, most of which are managed with a suppression objective, to pre-Euroamerican settlement fires– modern,

249 suppression fires are smaller on average compared to the natural range of variation, but modern large fires are

250 much larger compared to the natural range of variation (Safford and Stevens 2017; Miller and Safford 2017).

251 Fire effects of escaped suppression fires were more departed from effects of WFU fires for the same burn

252 duration (Figure 5). However, fire effects of suppression fires that succumbed to early suppression efforts

253 were generally beneficial when considering WFU fires as a reference for beneficial effects. We cannot use

254 our data to make a rigorous counterfactual prediction of how suppression fires that didn't survive initial

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²⁵⁵ attack may have ultimately influenced the overall distribution of fire effects if they had been allowed to burn.

²⁵⁶ However, the detected selection by suppression effect for extreme burning conditions suggests that, had these

²⁵⁷ fires burned for longer, they may have contributed to shifting the average fire effects of all suppression fires

²⁵⁸ closer to the observed fire effects for WFU fires (Figure 5).

²⁵⁹ **No detected effect of windspeed**

²⁶⁰ Surprisingly, we found no selection pressure on wind speed by early suppression efforts (Figure 3). Abatzoglou

²⁶¹ *et al.* (2018) found across the U.S. that greater wind speeds during the first two days after human-caused

²⁶² ignitions increased the likelihood that fires would grow large. Wind speed may not reflect a limiting factor in

²⁶³ early suppression efforts in this region compared to fuel continuity and fuel moisture as measured by energy

²⁶⁴ release component. Alternatively, the gridMET-derived wind speed in our study may be less representative

²⁶⁵ of how wind affects fire behavior in this particular system. That is, wind may play a strong role, but the

²⁶⁶ mountainous terrain may create more local wind conditions than can be captured by the gridMET product,

²⁶⁷ despite its high spatial and temporal resolution (Abatzoglou 2013).

²⁶⁸ **Selection beyond initial attack**

²⁶⁹ We demonstrated a strong directional effect of selection imposed by initial suppression efforts on the initial

²⁷⁰ burning conditions of wildfires in the Sierra yellow pine/mixed-conifer system. Fire effects arise from the

²⁷¹ confluence of fuel, weather, and topographical conditions in space and time, and thus our event-level (i.e.,

²⁷² per fire) measurements of burning conditions may mask some of this selection effect. We suspect this effect

²⁷³ may be even stronger if weather conditions were known at finer temporal scales and fuel conditions were

²⁷⁴ known at finer spatial scales. While we have shown a selection effect of suppression during an initial attack

²⁷⁵ response, the effect may not be limited to soon after ignition (i.e., the scope of our investigation). Whether

²⁷⁶ a similar selection by suppression effect would materialize beyond initial attack and *throughout* the course

²⁷⁷ of a fire's burn duration depends on the extent to which success in suppression is contingent upon fuel

²⁷⁸ and weather conditions cooperating with firefighting efforts. That is, preventing additional fire effects to

²⁷⁹ the landscape by extinguishing fires as soon as the weather or fuel conditions become more conducive to

²⁸⁰ suppression amounts to imposing a similar selection effect that we detected during initial attack suppression.

²⁸¹ Finer spatial and temporal resolution of burning conditions (e.g., daily climate variables paired with daily fire

²⁸² spread/severity maps) throughout the course of wildfires may help tune our understanding of how management

²⁸³ efforts can select for burning conditions. For example, aggressive suppression management may impose

²⁸⁴ the strongest selection pressure on fuel and climate conditions from the day of or the day prior to the fire.

²⁸⁵ Alternatively, WFU management may impose the strongest selection pressure on *future* fuel and climate

286 conditions, which would reflect managers' efforts to guide the spread of fire into forecasted spatiotemporally
287 favorable intersections of fuel and climate.

Positive feedbacks

289 Selection for more extreme burning conditions favors future extreme burning conditions in positive feedbacks
290 of fuel and climate. Homogeneous local forest structure makes this system more likely to burn at high severity,
291 with complete or nearly-complete mortality of overstory vegetation (Koontz *et al.* 2019b). While some
292 high severity fire is expected (Safford and Stevens 2017), elevated average levels and increased continuity of
293 high-severity fire and make it more likely that regenerating vegetation and future fuel structure will also
294 be homogeneous (North *et al.* 2009; Coppoletta *et al.* 2016; Miller and Safford 2017; Stevens *et al.* 2017).
295 On longer time scales, the selection by suppression for hotter/drier average burning conditions may favor
296 high-severity fire in the short run (Fried *et al.* 2004; Koontz *et al.* 2019b), and may compromise forest
297 recovery (Young *et al.* 2019), carbon stock stability (Earles *et al.* 2014), and carbon sequestration (Millar
298 and Stephenson 2015) in the long run. A reduction in carbon sequestration capacity contributes to climate
299 forcing, which is likely to perpetuate hot, dry conditions in California (Diffenbaugh *et al.* 2015; Mann and
300 Gleick 2015). Ongoing selection for extreme burning conditions by fire suppression therefore creates positive
301 feedbacks of burning conditions that are unfavorable to yellow pine/mixed-conifer forest persistence.

A selection effect on WFU fires, too

303 A conceptually similar, but directionally opposite selection pressure may bias the distribution of fire effects
304 that arise from wildfire use fires. The decision to let a candidate WFU ignition continue to burn as a WFU
305 fire when conditions are more mild selects for a distribution of fire effects that arises from these more mild
306 burning conditions. As with the selection by suppression, the distribution of fire effects of WFU fires will
307 be dictated by the fires whose fire effects were allowed to materialize (i.e., by deciding to let them burn in
308 the WFU case versus being unable to stop them from burning in the case of suppression fires that escape
309 initial attack). The fire effects from WFU fires are largely seen as beneficial, but their average effect should
310 be considered as a better-than-average scenario for reintroducing fire to previously fire-suppressed forests
311 compared to the expectations of fire effects if all candidate WFU ignitions were allowed to burn. Similarly,
312 the fire effects of suppression fires that escape initial attack are largely seen as negative, but their average
313 effect should be treated as a worse-than-average scenario for reintroducing fire to the landscape compared to
314 fire effects that would arise from letting all ignitions run their course.

315 The selection bias in both cases (suppression and WFU fires) reflects the social element in the coupled

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social-ecological system of wildfire and wildfire management. Management decisions, even as a fire burns, can hold great sway over ultimate fire effects to the forest system in their capacity to impose selection on the distribution of burning conditions. This strong influence of on-the-ground management decision-making points to the immense value of reducing barriers to implementing more wildfire use fires through increased training and reorganized incentive structures (Doane *et al.* 2006).

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