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High-severity burned area and proportion exceed historic conditions in Sierra Nevada, California, and adjacent ranges

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Peer reviewed

1	High severity burn area and proportion exceed historic conditions in forests of Sierra Nevada and
2	adjacent ranges (USA)
3	
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15	Open Research Statement: All data used in this manuscript will be made publicly available in the
16	Dryad Repository, of which the University of California is a member.
17	
18	ABSTRACT
19	Although fire is a fundamental ecological process in western North American forests, climate
20	warming and accumulating forest fuels due to fire suppression have led to wildfires that burn at
21	high severity across larger fractions of their footprint than were historically typical. These trends
22	have spiked upwards in recent years and are particularly pronounced in the Sierra Nevada-
23	southern Cascades ecoregion of California, USA and neighboring states. We assessed annual

area burned and percentage of area burned at high and low-to-moderate severity for seven major 24 forest types in this region from 1984 to 2020. We compared values for this period against 25 estimates for the pre-Euro-American settlement (EAS) period prior to 1850 and against a 26 previous study of trends from 1984-2009. Our results show that total average annual area burned 27 remained below pre-EAS levels, but that gap is decreasing (i.e., c. 14% of pre-EAS for 1984-28 29 2009, but 39% for 2010-2020 [including c. 150% in 2020]). Although average annual area burned has remained low compared to pre-EAS, both the average annual *area* burned at high 30 31 severity and the *percentage* of wildfire area burned at high severity have increased rapidly. The 32 percentage of area burned at high severity – which was already above pre-EAS average for the 1984-2009 period – has continued to rise for five of seven forest types. Notably, between 2010 33 and 2020, the average annual area burned at high severity exceeded the pre-EAS average for the 34 first time on record. By contrast, percentage of area that burned at low-to-moderate severity 35 decreased, particularly in the lower elevation oak and mixed conifer forest types. These findings 36 37 underline how forests historically adapted to frequent low-to-moderate severity fire are being reshaped by novel proportions and extents of high severity burning. The shift toward a high 38 severity-dominated fire regime is associated with ecological disruptions, including changes in 39 40 forest structure, species composition, carbon storage, wildlife habitat, ecosystem services, and resilience. Our results underscore the importance of finding a better balance between the current 41 42 management focus on fire suppression and one that puts greater emphasis on proactive fuel 43 reduction and increased forest resilience to climate change and ecological disturbance. 44 KEY WORDS: annual area burned; Cascade Range; fire ecology; fire regime; mixed conifer; 45 natural range of variation (NRV); Sierra Nevada; wildfire; yellow pine.

46

47 INTRODUCTION

Fire is a fundamental ecological process that has shaped the forests of western North America for 48 millions of years (Keeley and Safford 2016). The range of forest types found in this region is 49 related to the interactions of multiple factors, including climate, topography, species pool, 50 productivity, and disturbance history. These factors influence and are influenced by the fire 51 52 regime, which is defined by long-term temporal, spatial and fire intensity patterns of burning that typify an ecosystem and shape its composition, structure, and function (Agee 1993, van 53 54 Wagtendonk et al. 2018b, Miller and Safford 2020). Over the past century, however, fire regimes 55 in many western North American forests have departed from their natural range of variation. These modern changes have been driven largely by anthropogenic factors -e.g., halting of 56 Native American burning, adoption of fire suppression policies, timber extraction and forest 57 management practices, changing ignition patterns, and climate change – that have altered the 58 59 way fire interacts with forests (Abatzoglou and Williams 2016, Klimaszewski-Patterson and 60 Mensing 2016, Stephens et al. 2016, Balch et al. 2017, Parks et al. 2018a). A century of fire exclusion in the western United States has led to changes in fire 61 frequency and burn severity, two key components of the fire regime (Mallek et al. 2013, Safford 62 63 and Van de Water 2014, Steel et al. 2015, Parks et al. 2018b). The reduction or removal of regular fire has caused significant changes in forest structure, composition, diversity, and 64 65 function. For example, changes in forest fire regimes have promoted shifts in forest stand 66 density, fuel loading and continuity, and habitat heterogeneity (Johnstone et al. 2010, Hanberry 2014, Cassell et al. 2019, Stevens et al. 2019), and such shifts may be exacerbated by climate 67 68 warming (van Mantgem et al. 2013, Halofsky et al. 2020). In yellow pine (i.e., ponderosa pine 69 and/or Jeffrey pine; *Pinus ponderosa*, *P. jeffreyi*) and mixed conifer forests (the above species,

70	plus, among other species, sugar pine [P. lambertiana], incense cedar [Calocedrus decurrens]
71	and white fir [Abies concolor]), wildfires have grown in size and are more likely to include
72	larger contiguous patches of high severity burning than fires that burned prior to the application
73	of fire exclusion policies (Steel et al. 2018). Major changes in the yellow pine and mixed conifer
74	fire regimes have negatively impacted forest resilience, tree regeneration, species distributions
75	(Miller et al. 2009b, Keeley and Syphard 2016, Welch et al. 2016, Boisramé et al. 2017, Thorne
76	et al. 2017, Steel et al. 2018), threatened and endangered animal populations (Blomdahl et al.
77	2019, Jones et al. 2020), plant species diversity (Richter et al. 2019, Miller and Safford 2020),
78	and ecosystem services (Wu and Kim 2013, Richter et al. 2019, Rakhmatulina et al. 2020).
79	To better understand how wildfire patterns in the Western U.S. have been changing,
80	Mallek et al. (2013) compared modern vs. historical patterns of annual area burned and wildfire
81	severity in the Sierra Nevada-Southern Cascades ecoregion of eastern and northeastern
82	California and neighboring states. For the purposes of this study, "historical" refers to the time
83	before significant Euro-American settlement ("pre-EAS") of the study region, i.e., prior to
84	~1850. We also use the natural range of variation (NRV) to refer to the forest structure and
85	composition that existed pre-EAS, as defined by Safford and Stevens (2017; NRV as we use it
86	includes the contributions of Native Americans to the fire regime). Mallek et al. (2013) found
87	that while overall annual area burned in the period from 1984 to 2009 was only about 14% of
88	what burned in an average pre-EAS year (Stephens et al. 2007), the percentage of area burning at
89	high severity was much higher in 1984-2009 (29% area-weighted average vs c. 7% pre-EAS).
90	Another important finding of the study was that differences between the 1984-2009 and pre-EAS
91	periods depended on the forest type in question. For example, low- to mid-elevation forest types
92	(i.e., oak woodlands, yellow pine, mixed conifer) were burning much less frequently than under

pre-EAS conditions, but at much greater severity when they burned. By contrast, the authors
found that higher elevation forest types (i.e., red fir [*Abies magnifica*], lodgepole pine [*Pinus contorta*], subalpine forest), which have longer natural fire return intervals, experienced
relatively minor changes in fire frequency, and modern fire severity patterns were not
statistically discernable from pre-EAS patterns.

98 The annual area burned by wildfires in California has increased considerably since 2009 - the last data year considered by Mallek et al. (2013) - and nine of the ten largest fires in the 99 100 State's history have occurred since then (CalFire 2021b). Over the last decade, severe wildfires 101 have emitted hundreds of millions of Mg of carbon and other pollutants into the atmosphere (CARB 2020) and caused widespread ecological damage to forests, soils, and sensitive animal 102 habitat (Coppoletta et al. 2016, Jones et al. 2016, Welch et al. 2016, Abney et al. 2019, Dove et 103 104 al. 2020, Jones et al. 2020, Steel et al. In press). The Sierra Nevada-Southern Cascades ecoregion 105 (i.e., the study region; Fig. 1) has experienced similar trends in wildfire as California as a whole, 106 with regional variation driven by complex topography, prominent altitudinal gradients, and geographic clines in the distribution of climates and ecosystems (North et al. 2016, Safford et al. 107 2021). The recent large, high severity fires in the region, combined with the availability of eleven 108 109 additional years of fire severity data, led us to revisit and build on the analyses conducted by Mallek et al. (2013). 110

In this study, we provide an updated assessment of area burned and fire severity patterns for the Sierra Nevada-Southern Cascades region over the 37-year period from 1984-2020. Our goal is to provide the most current and refined assessment possible vis-à-vis changing fire regimes for resource managers who struggle to balance short-term conservation and risk aversion priorities with long-term considerations of ecosystem sustainability under rapid environmental

change. Specifically, we evaluate whether previously identified trends in burned area and fire 116 severity (Mallek et al. 2013) continue as before or whether they have slowed or accelerated. 117 Based on recent investigations (e.g., Steel et al. 2015, 2018; Safford and Stevens 2017) and 118 personal observations, we hypothesized that for the study region between 2010 and 2020: (1) 119 annual area burned would increase relative the 1984-2009 period, but would still lag behind 120 121 average annual area burned during the pre-EAS period; (2) the percentage of wildfire area burned at high severity would increase for all forest types, but proportionally more in low- and 122 123 middle-elevation forests, where forests have experienced greater departures from historical fire 124 return intervals due to a century of fire exclusion and climate warming; and (3) the average annual area burned at high severity (AAHS) would approach or exceed historical pre-EAS high 125 severity area in low- and middle-elevation forest types, but perhaps not in high-elevation forests, 126 given their longer fire return intervals and relative lack of fire for the last decade. 127

128

129 METHODS

130 *Study Area*

The study area comprises approximately 120,000 km² of the Sierra Nevada and Southern Cascade Mountain ranges and adjacent forested areas, and includes eleven National Forests and four National Parks (Fig. 1). This is the same study area used by Mallek et al. (2013) and Miller et al. (2009), and is based on the Sierra Nevada ecoregion as defined by the Sierra Nevada Ecosystem Project (SNEP 1996) and the Sierra Nevada Forest Plan Amendment (SNFPA; USDA 2004). The region stretches from Tehachapi Pass at the southern end of the Sierra Nevada to the California-Oregon border in the north, and from the Sierra Nevada Foothills on the eastern edge of California's Central Valley to the westernmost ranges of the Great Basin, including astrip of the Humboldt-Toiyabe National Forest in western Nevada.

140 Elevations in the study area range from 300 m above sea level along the western edge to >4000 m along the Sierra Nevada crest. The climate is mostly Mediterranean-type with warm, 141 dry summers and cool, wet winters. Vegetation in the study area is characterized by forests, 142 143 woodlands, shrublands, and grasslands, although the latter two are not analyzed here as our focus is on forested areas. Oak (Quercus spp.) woodlands dominate lower elevations along the western 144 boundary of the study area, transitioning to yellow pine (ponderosa pine [Pinus ponderosa] and 145 146 Jeffrey pine [P. jeffreyi]) and mixed conifer forests at higher elevations (Table 1). Red fir (Abies magnifica) dominated forests are found above about 1800 m and transition into lodgepole pine 147 (*P. contorta*) and different types of subalpine forest at the highest elevations. Pinyon pine 148 (mostly *P. monophylla*) and juniper (*Juniperus* spp.) woodlands occur at moderate elevations in 149 150 the north and east of the study area. Yellow pine-dominated forests are also found on the east 151 side of the study area, between about 1500 m and 2500 m elevation (Table 1; North et al. 2016, Safford et al. 2021). 152

153 Analyzed forest types and their areas are based on the LANDFIRE Biophysical Settings 154 (BpS) map (www.landfire.gov, v. 105, accessed 11/1/2019), which represents modeled potential natural vegetation incorporating climate, soils, topography and hypothetical pre-EAS (pre-1850) 155 156 fire regimes (Rollins 2009). BpS types were grouped into pre-settlement fire regime types 157 defined by Van de Water and Safford (2011) using crosswalks in that paper and Mallek et al. (2013). We analyzed the same seven forested pre-EAS fire regimes as Mallek et al. (2013) to 158 159 facilitate comparisons with that study: oak woodland (OW); dry mixed conifer (DMC); moist 160 mixed conifer (MMC); yellow pine (YP); red fir (RF); lodgepole pine (LP); and subalpine (SA).

While the BpS vegetation delimitations and pre-EAS fire regime estimates are the best-available for this analysis, we nevertheless stress that these parameters are based on a combination of incomplete data and historical reconstructions that necessarily mean that they should be viewed as approximations subject to refinement as new data and analytic methods become available.

165

166 Analysis

The data analyzed by Mallek et al. (2013) covered the time period from 1984 to 2009. In 167 168 this study, we used the most recent burn severity data available to consider eleven additional 169 years of wildfire extent and severity for the same region, extending the length of the period analyzed to the 37-years from 1984 to 2020. Wildfire perimeters and total annual area burned 170 (AAB) were obtained from the most recent version of the California Fire Perimeter database 171 (CalFire 2021a). The primary source of burn severity data for this analysis was the "Vegetation" 172 Burn Severity – 1984 to 2017" geospatial data layer (USDA 2018) developed by Region 5 173 174 (Pacific Southwest) of the United States Forest Service (henceforth "Forest Service"). For the 2018-2020 fire years, we estimated burn severity using Google Earth Engine following Parks et 175 al. (2018c, 2021). A comparison of 50 randomly selected fires from 1985-2017 showed high 176 177 similarity between the legacy and Earth Engine-derived severity estimates (R = 0.95; Figure S1). For both datasets, severity data were calculated from Landsat Thematic Mapper imagery using 178 179 the Relative differenced Normalized Burn Ratio (RdNBR) and were classified into severity 180 levels using previously field-calibrated thresholds (Miller and Thode 2007, Miller et al. 2009a). 181 The dataset includes the entire area of all wildfires ≥ 80 ha in size that occurred at least partially 182 on Forest Service lands or in Yosemite National Park in the study area, plus an incomplete 183 collection of fires <80 ha (see: Miller et al. 2009b, Miller and Safford 2012, Mallek et al. 2013).

We did not include Lassen or Sequoia-Kings Canyon National Parks because fire severity
mapping for fires <400 ha has not been carried out in these landscapes.

186 We used burn severity data to calculate hectares burned in four fire-severity classes (per Miller and Thode 2007) for each forest type for each year from 1984 to 2020. Like Mallek et al. 187 (2013), we condensed the severity data into two categories: 1) "annual area burned at low-to-188 189 moderate severity" (AALMS), a single category that combines classes I ("no change"), II ("low severity" = <25% tree mortality), and III ("moderate severity" = 25 to <95% tree mortality); and 190 2) "annual area burned at high severity" (AAHS), which represents class IV burned areas that 191 192 experienced stand-replacing fire, where tree mortality at the time of postfire imagery acquisition was >95% (Miller et al. 2009a). For all areas analyzed for severity, total annual area burned 193 (AAB) for a forest type was equal to AAHS plus AALMS. 194 For the pre-EAS burn data, we used the same numbers and methods used by Mallek et al. 195 (2013), with a few updates to the average fire rotation period based on new science (see below; 196 197 Table 2), defined as the number of years required to burn an area equal to the forest extent in question (Agee 1993). We used the pre-settlement fire regime types cross-walked from the 198

199 LANDFIRE BpS map (see above) and divided the total area of each type by its pre-EAS fire

200 rotation period (Table 2) to estimate average AAB_{Pre}. Thus, for an area, A, associated with a pre-

EAS fire regime rotation period of Y years, $AAB_{Pre} = A/Y ha \cdot yr^{-1}$

Whereas the burn severity class data for the modern period are imagery-based, our estimates of characteristic burn severity for the pre-EAS period were made from historical records, the scientific literature, and models. We started from Table 3 in Mallek et al. (2013) and consulted the literature for updated information. Based on new data summarized in Safford and Stevens (2017), we did not change the Mallek et al. (2013) estimates of characteristic burn severity levels for oak woodland, dry and moist mixed conifer, or yellow pine. However, we
adjusted the values for red fir, lodgepole pine, and subalpine forest based on new information
(Safford and Stevens 2017, van Wagtendonk et al. 2018a, Meyer and North 2019). These sources
yielded AAB_{Pre} and AAHS_{Pre}, from which we calculated AALMS_{Pre} (AAB minus AAHS),
percentage of area burned at high severity (PHS; = AAHS/AAB), and percentage of area burned
at low-to-moderate severity (PLMS; = AALMS/AAB).

As in Mallek et al. (2013), we intersected the SNFPA polygon with the LandFire BpS 213 214 raster data set (version 105) to define the major vegetation classes. We also added fire severity 215 data for a few fires that burned in the study area during the Mallek et al. time frame but were not analyzed for severity in that study. No areas outside of the study area polygon were analyzed or 216 reported, even if part of a given fire burned inside the boundary. As in Mallek et al. (2013), we 217 included all fires >80 ha that intersected both the study area polygon and the Forest Service or 218 219 Yosemite National Park jurisdictions, while those that did not were excluded from the severity 220 analysis (Figure 1). Sections of fires that fit these criteria but fell outside of the study area boundary were excluded from the analyses. Our data for the period 1984-2009 are nearly the 221 same as, though not identical to, those used by Mallek et al. (2013) because of subsequent 222 223 updates to the Forest Service fire severity database and our revised PHS estimates for the pre-EAS period for red fir, lodgepole, and subalpine forests. As in Mallek et. al. (2013), we included 224 225 all fires >80 ha that intersected both the study area polygon and the Forest Service or Yosemite 226 National Park jurisdictions, while those that did not were excluded from the analysis (Figure 1). 227

228 Trend Assessment

We used a Bayesian approach to assess trends in AAB, PHS, and PLMS for the full study 229 region and by forest type across the expanded modern period (1984-2020). For this assessment 230 we fit generalized linear models with year as the fixed effect of interest. Area response variables 231 were log-transformed and modeled using a Gaussian error structure. Proportion burned area 232 models utilized aggregated binomial regression and a logit link function with hectares of 233 234 AALMS or AAHS constituting "successes" and AAB constituting "trials" for a given year and forest type. In all models we included a first-order temporal auto-regressive term to account for 235 236 potential temporal auto-correlation.

Models were estimated using Hamiltonian Monte Carlo sampling in Stan via the BRMS package and program R (Bürkner 2017, R_Core_Team 2019, Stan_Development_Team 2019). We specified weakly regularizing priors to prevent model overfitting. Models were run with three chains, each for 3000 samples with a warmup of 1500. Trace-plots and R-hat values were assessed for proper mixing and model convergence.

242

243 RESULTS

Average annual area burned (AAB) during the 2010-2020 period – though still well below historic levels (AAB_{Pre}) – increased by more than 200% over the 1984-2009 period for all forest types combined. AAB₂₀₁₀₋₂₀₂₀ was especially impacted by the record-breaking 2020 wildfire season (Table 3, Fig. 2a, Safford et al. 2022), which contributed significantly to the large overall increases in AAB across the expanded modern period (1984-2020) for all forest types, individually and combined.

The average annual percent of area burned at high severity (PHS) increased for all forest types combined between the 1984-2009 and 2010-2020 periods (Table 3, Fig. 3). When these

252	two periods are considered together, PHS ₁₉₈₄₋₂₀₂₀ averaged 27% - almost four times the combined
253	PHS _{Pre} average of 7%. For some forest types, however, PHS did not increase from 1984-2009 to
254	2010-2020. For yellow pine, for example, PHS was virtually unchanged across the two modern
255	periods (though still much higher than pre-EAS values). PHS ₂₀₁₀₋₂₀₂₀ also decreased for
256	lodgepole and sub-alpine forests compared to PHS ₁₉₈₄₋₂₀₀₉ (Figs. 3, S2, S3). By contrast, PHS ₂₀₁₀₋
257	2020 trended noticeably upward for the oak woodland, dry and moist mixed-conifer and red fir
258	forests. The complement of PHS, PLMS (percentage of area burned at low-to-moderate
259	severity), showed a decreasing trend overall from 1984-2009 to 2010-2020, with yellow pine,
260	lodgepole pine and subalpine forests as individual exceptions.
261	The average annual area burned at low-to-moderate severity (AALMS) increased since
262	2009 across all forest types, but remained well below historical (AALMSPre) levels. Notably, for
263	2010-2020 the average annual area burned at high severity (AAHS) exceeded pre-EAS levels for
264	the first time on record (Fig. 4). These trends are visible for all forest types combined, as well as
265	for the dry and moist mixed conifer, yellow pine, and red fir forest types separately (Table 3, Fig.
266	4b).
267	For the 2010-2020 period, all forest types showed appreciable increases in AAB
268	compared to 1984-2009 (average increase: 410%; range: 56 – 905%, Table 3). AAB increased
269	from 13.6% of AAB_{Pre} during 1984-2009 to 39% of AAB_{Pre} during 2010-2020 (including c.
270	150% of AAB _{Pre} in 2020 alone; Table 3). For the expanded modern period, 1984-2020, annual
271	area burned averaged 20.6% of AAB_{Pre} across forest types and ranged from 14.6% (oak
272	woodland) to 34.8% (red fir). Thus, despite recent increases, average annual area burned
273	continues to be less than half of AAB _{Pre} , due to an ongoing deficit in low-to-moderate severity
274	fire (Fig. 4a).

A comparison of modeled trends across the 1984-2020 period for burned area and burn 275 severity revealed similarities and differences among forest types (Figs. 5, S3). For example, 276 277 AAB₁₉₈₄₋₂₀₂₀ and AAHS₁₉₈₄₋₂₀₂₀ showed positive trends over time across all forest types, though the amount of increase varied in absolute and relative terms. Subalpine, lodgepole pine, and 278 moist mixed conifer – in that order – showed the most robust increases in $AAB_{1984-2020}$, while dry 279 280 mixed conifer, moist mixed conifer, and red fir had the strongest positive trends in AAHS₁₉₈₄- $_{2020}$. For all forest types combined, PHS₁₉₈₄₋₂₀₂₀ trended positive for the expanded evaluation 281 282 period. The results for this trend and AAHS₁₉₈₄₋₂₀₂₀ were still positive and significant when the 283 2020 fire year was excluded. For PLMS₁₉₈₄₋₂₀₂₀, in terms of individual forest types, only yellow pine showed a convincingly stable trend; all other forest types showed decreasing trends. 284 285

286 DISCUSSION

Our findings support previous assessments of burned area and severity in California 287 288 (Miller et al. 2009b, Mallek et al. 2013, Miller and Safford 2012, Steel et al. 2015), but go further in demonstrating that high severity trends have surpassed historical rates and have 289 stepped up markedly since 2009. While part of this jump is due to the record 2020 fire year 290 291 (Safford et al. 2022), the increases in high severity fire in recent years are remarkable even when 2020 is not considered. The most salient results of our assessment are that: (1) average annual 292 293 area burned (AAB₁₉₈₄₋₂₀₂₀) remains well below pre-EAS averages, although the disparity is 294 decreasing; (2) for the newly evaluated 2010-2020 period, average annual area burned at high severity (AAHS₂₀₁₀₋₂₀₂₀) exceeded AAHS_{Pre} for the first time on historical record, particularly in 295 296 low and middle elevation forest types; and (3) the percentage of area burned at high severity 297 during the expanded modern period (PHS₁₉₈₄₋₂₀₂₀) is well above pre-EAS levels and trending

upward for six of seven forest types analyzed (Fig. S2). Conversely, the percentage of area
burned at low-to-moderate severity (PLMS₁₉₈₄₋₂₀₂₀) shows a decreasing trend that adds to an
already gaping deficit in the type of burning that is fundamental to the conservation and
restoration of most of the Sierra Nevada-Southern Cascades forest base (van Wagtendonk et al.
2018b).

303 Our data show that the gap between AAB₁₉₈₄₋₂₀₂₀ and AAB_{Pre} is closing, due mainly to increases in the area burned at high severity. In California and adjoining western states, forest 304 305 types such as oak woodland and yellow pine-mixed conifer evolved under fire regimes 306 characterized by frequent, low-to-moderate severity burning (Agee 1993, Van Wagtendonk et al. 2018b, Safford et al. 2021). The dominant tree species in these forests are resistant to fire as 307 adults, with adaptations like thick bark, self-pruning of lower branches, thick cone scales, and 308 highly flammable needle cast that serves to reduce competition from seedlings and saplings 309 when it burns (Safford and Stevens 2017). Most of these species are not adapted to high severity 310 311 fire, however (Keeley and Safford 2016).

As a result of the relative increases in high severity fire and the concomitant reductions in 312 low-to-moderate severity fire, researchers have documented major ecological impacts on the 313 314 study region. These changes include: loss of carbon storage; increased plume emissions and decreased air quality; increased erosion; and adverse impacts on soil nutrients, microbial 315 316 processes and hydrology (Maestrini et al. 2017, Roche et al. 2018, Abney et al. 2019, Dove et al. 317 2020). Additionally, studies have shown that shifts in burning patterns correlate with failures in 318 conifer regeneration (Welch et al. 2016, Shive et al. 2018), changes in the balance of fire tolerant 319 and fire intolerant species (Stevens et al. 2015, White et al. 2016), negative impacts to overall 320 species diversity and to many plant and animal taxa (Miller et al. 2018, Blomdahl et al. 2019,

Dalrymple and Safford 2019, Richter et al. 2019, Steel et al. 2019, Jones et al. 2020, Steel et al. 321 2021), and vegetation type conversion (Webster and Halpern 2010, Collins et al. 2011, Stevens 322 et al. 2015, Coppoletta et al. 2016, Tepley et al. 2017, Coop et al. 2020, Dove et al. 2020). To 323 reverse these changes and restore the fire regime processes to which the dominant oak, yellow 324 pine and mixed conifer forest types are historically adapted, it will be necessary to substantially 325 326 increase the area and percentage of forest burned at low-to-moderate severity (Scholl and Taylor 327 2010, North et al. 2012, Safford and Van de Water 2014). Given the severity trends presented 328 here (and further explored in Safford et al. 2022), wildfire alone appears unlikely to produce the 329 kind of mixed-severity burning that historically characterized these forests. Instead, achieving these goals will likely require increased use of prescribed fire, wildfire managed for resource 330 benefit and/or other types of intentional fuel treatments. 331

Compared to lower elevation forests, red fir, lodgepole pine and subalpine forests – 332 characterized by patchy, often rocky landscapes, slow rates of growth and fuel accumulation, and 333 334 colder, shorter fire seasons – have infrequent fires and higher interannual variability in area burned, making trends harder to discern (van Wagtendonk et al. 2018b, Meyer and North 2019). 335 The natural range of variation is also more difficult to define for these forest types because they 336 337 have longer fire return intervals and historical data are harder to find the further one goes back in time. That said, there were two findings in our results for these forest types that we can interpret. 338 339 First, while red fir forests experienced a 74% increase in PHS between 1984-2009 and 2010-340 2020, lodgepole pine and subalpine forests averaged decreases in PHS between these two periods (-16% and -46%, respectively). Second, although average annual area burned in 2010 to 2020 341 342 was lower than AAB_{Pre} for all forest types, the deficit decreased markedly in these high elevation 343 forests, including roughly 10-fold increases in average annual area burned for lodgepole pine and

subalpine forests over AAB₁₉₈₄₋₂₀₀₉. These findings suggest that fire suppression has less of an 344 impact on historical/NRV fire severity and burn patterns at the highest elevations, especially 345 where lodgepole pine and subalpine forests are typically found. We consider the most 346 compelling explanation to be because the lack of fire over the last century represents a smaller 347 departure from the pre-EAS fire return intervals compared to forest types adapted to more 348 349 frequent fire (Safford et al. 2012b, Safford and Van de Water 2014). Another contributing factor is likely that fire suppression is implemented less in high elevation forests due to reduced access, 350 351 low density of human assets and fire management policies that are more tolerant of naturally 352 ignited fire for ecological benefit (van Wagtendonk 2007).

When comparing current burn trends to historical ones, it is important to consider the data 353 accuracy for both time periods. California's fire perimeter dataset is highly accurate after 1950 354 and the Landsat imagery that makes complete region-wide fire severity mapping possible has 355 356 been available since 1984 (Miller et al. 2009b). Moreover, the availability of severity atlases and 357 statistical models that relate severity maps to ground-based measurements is constantly expanding. The Forest Service RdNBR-based dataset for California is likely the most 358 trustworthy in the US – it has been extensively ground-validated and calibrated, many smaller 359 360 fires are included in the dataset, and fire severity classifications use objective thresholds that allow translation of fire effects into biomass loss, permitting comparisons across fires and years 361 362 (see e.g., Miller and Thode 2007, Safford et al. 2008, Miller et al. 2009a, Miller et al. 2016). 363 Further, the development of partially automated approaches (e.g., using Google Earth Engine) allow for consistent and comprehensive fire severity estimates across broad geographies (Parks 364 365 et al. 2018, 2019).

In contrast, it is difficult to estimate historical fire severity and rotation periods with high 366 precision because they are (a) variable by nature and (b) based on patchy reconstruction 367 estimates that only get more difficult to piece together the further back one goes in time. We 368 used recent studies (Mallek et al. 2013, Safford and Van de Water 2014) and natural range of 369 variation studies (Safford and Stevens 2017, Meyer and North 2019) to inform our estimates 370 371 because they represent thoroughly researched, best-available inferences that combine historical data, modern reference sites, current research, and model-based assessments of both the study 372 373 system in question and adjoining analogous systems. We do not discount the unavoidable 374 imprecision that comes with reconstructing historical fire return intervals and severity patterns across time spans for which data are largely absent. Nevertheless, we believe a more pressing 375 challenge facing future studies may be to determine the likely future range of variability under 376 emerging climatic conditions (Wiens et al. 2012). 377

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379 MANAGEMENT IMPLICATIONS

Our findings have important implications for fire and forest management, policy, and 380 conservation in and around the study region. First, although it has been widely known for more 381 382 than 50 years that fire exclusion in western US forests is a major driver of ecosystem and fire regime change, many federal and state agencies persist in suppressing almost all fires (Calkin et 383 384 al. 2005, Stephens et al. 2016). Wildfire suppression will continue to be necessary to protect 385 human life, property and other important assets, but in fire-adapted landscapes it should be 386 considered as only one of many tools in the management toolkit. Continued focus on reducing 387 burned area – even in ecosystems where the principal ecological missing-link is fire, such as oak 388 woodlands, yellow pine and mixed conifer forests – will not address the urgent need to minimize the ecologically harmful impacts of fire (Stephens et al. 2016, Moreira et al. 2020, Safford et al.2022).

391 The disconnect between fire management and resource management was the chief driver of the switch from blanket fire suppression to multipurpose fire management that was made in 392 US federal agencies in the late 1960s and early 1970s (Stephens and Ruth 2005), as well as in the 393 394 2009 update to US federal fire management policy that permitted all wildfires to be managed for suppression and/or resource benefit (USDA-USDOI 2009). However, the proportion of the 395 396 Forest Service budget that goes to wildfire suppression-related activities rose from 16% in 1995 397 to 52% in 2015 (Stephens et al. 2016), and exceeds 65% today. As North et al. (2015) note, myopic focus on short-term fire management results not from policy constraints but from 398 "entrenched agency disincentives to working with fire." These disincentives relate to nuances of 399 budget allocation, concerns about assets at risk, smoke production, politics, liability, and public 400 perception of all fire as bad (Calkin et al. 2015). Whatever the drivers, as fires grow larger and 401 402 spread more rapidly, increasingly large portions of the annual budgets of federal resource management agencies are diverted to putting out fires, siphoning already scarce funding from 403 proactive ecosystem management and restoration activities (including fuel reduction) and 404 405 paradoxically increasing the potential for severe fires in the future as fuels continue to accumulate and the climate continues to warm (Carroll et al. 2007, Calkin et al. 2015, Stephens 406 407 et al. 2016, Moreira et al. 2020).

The fire-climate modeling literature (e.g., Dettinger et al. 2018, Restaino and Safford 2018) also projects increases in annual area burned that are consistent with our findings. These trends have generated excited headlines that decry a "climate reckoning in fire-stricken California" (NY Times, September 10, 2020) and warn of wildfires in the West "spread[ing] like

the plague" (Wall Street Journal, September 8, 2020). However, increasing burned area, the most 412 often cited measure of calamity, is only an ecological concern where annual burning exceeds the 413 NRV, routinely and over the long-term. The 2020 wildfire season was the only year in our study 414 period that came close to being comparable in burned area to the pre-Euro-American settlement 415 average. That said, there *are* ecosystems in California and neighboring states where annual 416 417 burned area is unsustainably high by ecological standards. These are primarily sagebrush and related ecosystems in the Great Basin and chaparral and sage scrub in central and southern 418 419 California, where the problem is driven by highly flammable invasive annual grasses, and in 420 chaparral, a surfeit of human ignitions (Safford et al. 2018, Safford et al. 2021). In these places, fire suppression is both ecologically justified and crucial. 421

For the forest types we analyzed, however, the issue is not too much burning but too 422 much of *the wrong kind* of burning. The tendency of modern forest fires that escape initial attack 423 to burn large areas at high severity is driven by (1) unnaturally high fuel loadings and (2) 424 425 weather conditions that reflect a steadily warming climate (Abatzoglou and Williams 2016, Keeley and Safford 2016, Parks et al. 2018b, Safford et al. 2021, 2022). For the most part, 426 increased investment in fire suppression is a short-term fix that fails to resolve these issues and, 427 428 when "successfully" implemented, extends the period of fuel accumulation. While essential for the protection of life and property in the wildland-urban interface, and thus of relevance to any 429 430 comprehensive solution to wildfire (Schwartz and Syphard 2021), fire suppression of natural 431 ignitions can have an aggravating effect when applied to forest types adapted to frequent fire 432 (Moreira et al. 2020). By contrast, climate change mitigation will be fundamentally important in 433 the long-term, but will not address the immediate need to reduce fuels in erstwhile frequent-fire 434 forest types (e.g., oak woodland, yellow pine, mixed conifer) where fire regime changes and

ecologically damaging fires have been most pronounced (Steel et al. 2015). Instead, this
objective may be accomplished through strategic expansion of active fuels reduction, enhanced
application of prescribed fire, and increased management of wildfires for ecological purposes
(i.e., "resource benefits"), alone or in combination (North et al. 2012, North et al. 2015, Stephens
et al. 2021).

440 While by no means the definitive source for setting fire-related management targets, parameters provide forest managers with a useful template for considering burn frequency and 441 442 severity objectives in the context of historical forest structure and composition. By comparing a contemporary forest to its NRV, managers can assess whether restoration to such standards is (a) 443 appropriate and (b) feasible based on how much a forest resembles or is departed from the 444 conditions under which it presumably functioned before Euro-American settlement (Manley et 445 al. 1995, Landres et al. 1999, Wiens et al. 2012). In the case of yellow pine and mixed conifer 446 447 forests in our study region, for example, comparisons of contemporary forest stands with NRV 448 reveal forests with tree densities that are 2-4x higher (or more) than before EAS, average tree diameters about half of their historical norms, higher and more continuous canopy cover, and 70-449 100% increases in surface fuel loadings – changes that suggest modern stands are more fuels-450 451 limited than ignition-limited (Safford and Stevens 2017).

Because anthropogenic warming is leading us away from the climatic conditions that characterized the pre-EAS/NRV period, it has been suggested that NRV-based targets should be applied cautiously (Millar et al. 2007). However, Safford et al. (2012a, 2012b) point out that under shifting environmental baselines, NRV conditions retain their value, especially where they are interpreted as management reference points rather than endpoints, and where they are used to better understand the mechanisms of change. Research suggests that future forests in the study

region will support lower tree densities and biomass than under current or pre-EAS conditions 458 (Lenihan et al. 2003, Safford and Stevens 2017, Stanke et al. 2021, North et al. 2022). If so, 459 managers could use NRV estimates as a reference point from which to set new targets for forest 460 resilience based on how much current and NRV conditions differ. Substantiation for the value of 461 the NRV in the study area is also found in recent research into the fire responses of key wildlife 462 463 indicator species (California spotted owl [Strix occidentalis occidentalis], Pacific fisher [Pekania *pennanti*], and black-backed woodpecker [*Picoides arcticus*]), whose nesting and foraging 464 behaviors show strong links to pre-EAS ranges of variation in fire severity and high severity 465 patch size (Safford and Stevens 2017, Blomdahl et al. 2019, Stillman et al. 2019, Jones et al. 466 2020, Kramer et al. 2021). Thus, we see a natural synergy between (a) studies such as this one 467 that provide a multi-decadal perspective on how fire patterns are changing across a cohesive 468 landscape and (b) NRV-type assessments that provide managers and researchers with an 469 ecologically meaningful context in which to consider the implications of those changes and what 470 471 actions they might implement in response.

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759 TABLES AND FIGURES

760 A)

Forest type (code)	Dominant species	Average	Extent (ha)	Burned area mapped for		
		elevation (m)		severity (1984-2020)		
Oak woodland (OW)	QUDO, QUWI, PISA	756	959,252	275,744		
Dry mixed conifer (DMC)	PIPO, PILA, CADE,	1121	737 031	267 624		
Dry mixed conner (Divic)	ABCO, QUKE	1121	151,751	207,024		
Moist mixed conifer (MMC)	ABCO, PSME, PILA,	1500	1 372 110	112 750		
worst mixed conner (wiwie)	CADE, SEGI	1390	1,372,110	442,739		
Yellow pine (YP)	PIJE, PIPO, QUKE	1,714	1,550,530	442,701		
Red fir (RF)	ABMA, PIMO	2335	1,026,116	169,204		
Lodgepole pine (LP)	PICO	2786	111,178	9,640		
Subalning (SA)	PIAL, PIMO, PIFL,	2162	264 175	6 202		
Subalphie (SA)	PICO, TSME	5105	204,175	0,592		

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762 B)

Acronym	Explanation
AAB	Annual area burned (all severity classes)
AAHS	Annual area burned at high severity (class IV)
AALMS	Annual area burned at low-to-moderate severity (classes I-III)
PHS	Percentage of burned area burned at high severity
PLMS	Percentage of burned area burned at low-to-moderate severity
EAS	Euro-American settlement (~1850)
Pre [as subscript]	Refers to Pre-EAS, i.e., before ~1850

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Table 1. A) Forest types considered in this study. Dominant tree species that characterize each

type are listed using the following abbreviations: ABCO: *Abies concolor*; *ABMA: A. magnifica;*

- 766 CADE: Calocedrus decurrens; PIAL: Pinus albicaulis; PICO: P. contorta ssp. murrayana;
- 767 PIFL: P. flexilis; PIJE: P. jeffreyi; PILA: P. lambertiana; PIMO: P. monticola; PIPO: P.
- 768 ponderosa; PISA: P. sabiniana; PSME: Pseudotsuga menziesii; QUDO: Quercus douglasii;
- 769 *QUKE: Q. kelloggii; QUWI: Q. wislizenii; SEGI: Sequoiadendron giganteum; TSME: Tsuga*

- *mertensiana*. B) An explanation of the acronyms used for the variables and time periods
- 771 analyzed.

Forest	Fire rotation		PHS	Source
	Average		(%)	
Туре	(yrs) Range			Literature
OW	18	12-25	6	Mallek et al. 2013
DMC	23	11-34	6	Mallek et al. 2013, Safford & Stevens 2017
MMC	31	15-70	8	Mallek et al. 2013, Safford & Stevens 2017
YP	22	11-34	5	Mallek et al. 2013, Safford & Stevens 2017
RF	79	25-163	10	Miller et al. 2012, Mallek et al. 2013, Meyer and North 2019
LP	63	46-80	24	Mallek et al. 2013, Meyer and North 2019
SA	425 75-721		10	Mallek et al. 2013, Meyer and North 2019, van Wagtendonk et al. 2018a

Table 2. Estimated average Pre-Euro-American Settlement fire rotation period in years and
percent burned at high severity (PHS) for the forest types considered in this study. Estimates are

based on the average values for the range of numbers found in the corresponding published

scientific literature sources. Key to forest type codes found in Table 1.

AAB (ha)				PHS PLMS			AAHS (ha)			AALMS (ha)					
Туре	Pre	1984- 2009	2010- 2020	Pre	1984- 2009	2010- 2020	Pre	1984- 2009	2010- 2020	Pre	1984- 2009	2010- 2020	Pre	1984- 2009	2010- 2020
All	211,822	27,154	82,551	7%	29%	36%	93%	71%	64%	14,002	7,955	30,001	197,819	19,199	52,550
OW	51,168	6,387	9,972	6%	22%	32%	94%	78%	68%	3,275	1,421	3,189	47,893	4,966	6,783
DMC	31,461	3,947	15,001	6%	25%	43%	94%	75%	57%	1,903	986	6,411	29,558	2,960	8,590
MMC	44,076	5,328	27,657	8%	30%	37%	92%	70%	63%	3,658	1,600	10,172	40,418	3,728	17,485
YP	69,411	8,360	20,485	5%	42%	39%	95%	58%	61%	3,349	3,511	8,066	66,062	4,850	12,419
RF	13,132	3,014	8,258	10%	14%	24%	90%	86%	76%	1,313	411	1,951	11,819	2,603	6,307
LP	1,758	71	710	24%	30%	26%	76%	70%	74%	422	21	182	1,336	49	527
SA	816	47	469	10%	12%	6%	90%	88%	94%	82	6	30	734	42	439

Table 3. Comparison of average annual burned area and percentage burned at different severity classes for the study area by forest

type and time period. Total annual area burned (AAB) in hectares is the sum of annual area burned at high severity (AAHS) and

- annual area burned at low-to-moderate severity (AALMS) severity. AAHS/AAB is the percentage burned at high severity (PHS) and
- 781 AALMS/AAB is the percentage burned at low-to-moderate severity (PLMS). Average annual percentage values listed are not

weighted by annual burned area. "Pre" refers to the pre-Euro-American Settlement before 1850. Forest type codes as in Table 1.







A)





- on previous literature. The solid upsloping trend line shows the fitted linear model from this
- study with log area as the response variable and time as the predictor variable with an
- autoregressive term. B) AAB by major forest type for the periods 1984-2009 (white bars) and
- 801 2010-2020 (black bars) as a percentage of AAB_{Pre} (left axis). White and black circles show
- AAB in hectares for the same two periods, respectively (right axis, note log scale).





Figure 3. Burn severity trends as a percentage of total area burned averaged across years for 804 three time periods: prior to ~1850 (Pre); 1984 - 2009 (84-09); and 2010 - 2020 (10-20). Blue 805 806 bars are percentage burned at low-to-moderate severity (PLMS); orange bars are percentage burned at high severity (PHS). Cumulative data for all forest types combined are indicated by 807 "All" and separated with a vertical dashed line. Slight differences between 1984-2009 values and 808 809 values in Mallek et al. (2013) are due (1) to addition of pre-2010 fires to the burn severity dataset after 2013, and (2) to changes in pre-EAS fire severity due to new information (see Table 3). See 810 Methods for details. Forest type codes as in Table 1. 811



- Figure 4. Comparison of average annual area burned (hectares) in the Sierra Nevada-Southern
- 835 Cascades study region by forest type for A) low-to-moderate severity fire (AALMS) and B) high
- severity fire (AAHS). Gray bars are estimates for pre-Euro-American settlement ("Pre"); blue
- bars are for the period 1984-2009; and orange bars are for the period 2010-2020. Forest type
- codes as in Table 1. Error bars based on standard error of the mean.



Figure 5. Standardized trend estimates by forest type for wildfire burned area and severity in the Sierra Nevada-Southern Cascades from 1984 to 2020. Trends were derived using generalized linear models and are for total annual area burned (AAB) and its high severity (AAHS) and lowto-moderate severity (AALMS) components, together with estimates in trends for percentage of area burned at high (PHS) and low-to-moderate severity (PLMS). Estimates to the right and left of the dashed lines indicate increasing and decreasing trends with time, respectively. Forest type codes as in Table 1.

847 SUPPORTING INFORMATION



848

Figure S1. A comparison of burned area and percent high severity for 50 randomly selected

850 1985-2017 wildfires calculated using the legacy ("Miller method") and Google Earth Engine-









Figure S2. Temporal trends of high severity (PHS) burned area as a percentage of total area
burned by wildfire for different forest types of the Sierra Nevada-Southern Cascades region.
Linear regression trendlines for the period 1984 to 2009 are shown in blue, while those for the
period 2010 to 2020 are shown in orange. The combined period trendline is shown by dotted
black lines.





Figure S3. Generalized linear model (GLM) trend lines and credibility intervals (gray area) by
forest type for percent of area burned annually by high severity wildfire (PHS) in the Sierra
Nevada-Southern Cascades study region for the extended modern period (1984-2017) considered
in this study. Forest type codes as in Table 1.